

Novas Aplicações das Matemáticas na Indústria

Osamu Saeki

Institute of Mathematics for Industry
Kyushu University
Japan



I. Institute of Mathematics for Industry (IMI)

**II. Application of Singularity Theory to
Visualization of Big Data**

Policy Study in 2006 by the Japanese Government

Europe and USA:

65% of workers in R&D departments in private companies have **Mathematics** as background.

Japan: only **26%**

This shortage and nearly **40%** gap must be overcome.

In Japan, **Pure Mathematics** have been studied much more than **Applied Mathematics**.

What is Mathematics-for-Industry ?

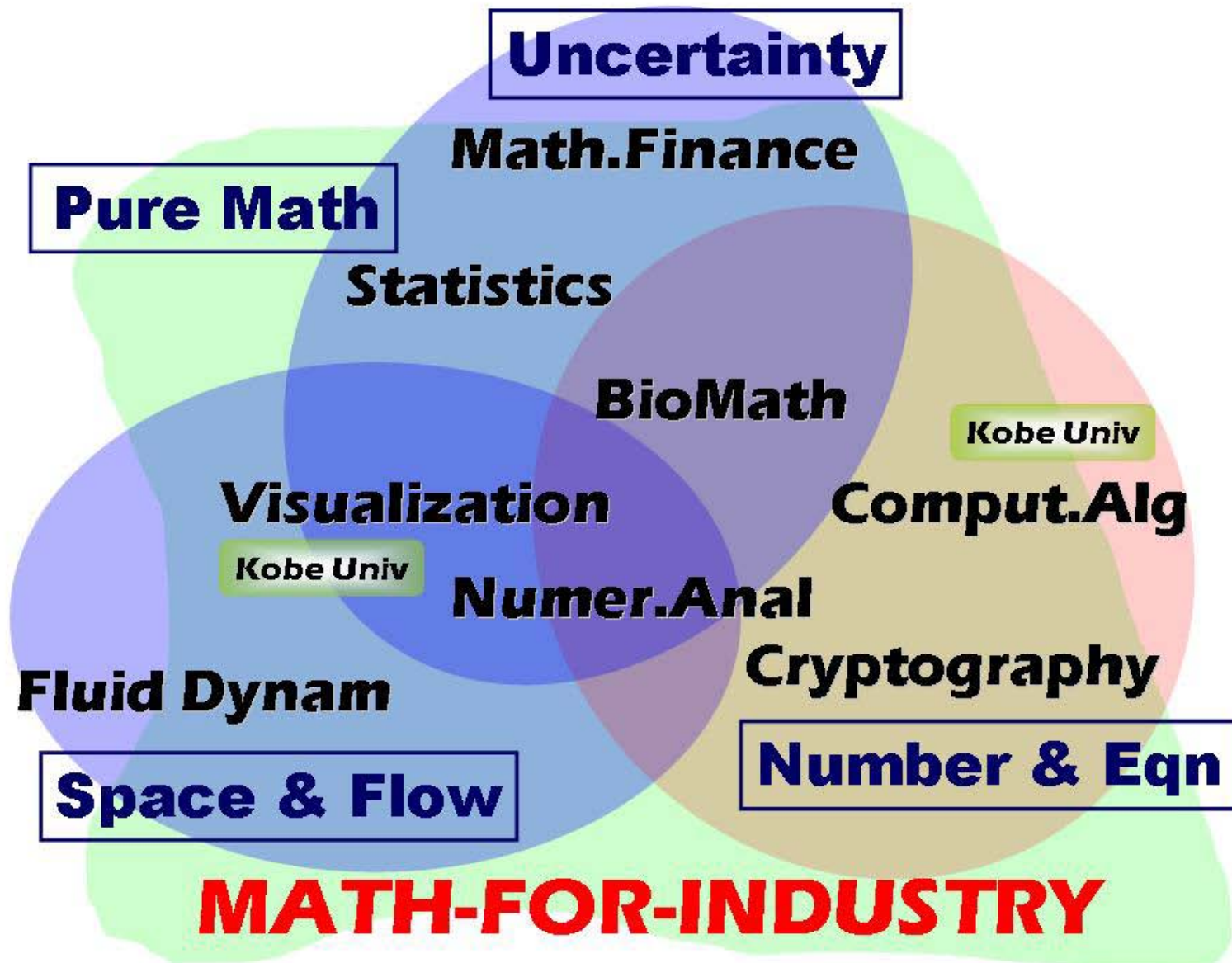
- Mathematics-for-Industry (**MI**, for short) is a **new research area** that will provide a foundation for creating future technologies. **MI** responds to the industrial needs by reorganizing and merging **pure and applied mathematics**.
- Main purpose of our program is to perform the **education and research** activities in **MI**.

<http://gcoe-mi.jp/>

Global **C**enter **O**f **E**xcellence
Program (like CEPID-FAPESP)
April 2008- March 2013



Disciplines covered by MI



Forum “Math-for-Industry”

FMI 2009 Fukuoka November 9-13 Casimir Force, Casimir Operators and the Riemann Hypothesis



FMI 2010 Fukuoka October 21-23 Information Security, Visualization, and Inverse Problems, on the basis of Optimization Techniques



FMI 2011 Honolulu October 24- 28 TSUNAMI - Mathematical Modelling - Using Mathematics for Natural Disaster Prediction, Recovery and Provision for the Future -



FMI 2012 Fukuoka October 22-26 Information Recovery and Discovery



Study Group Workshops (2010, 2011, 2012)

3rd SGW: July 25-27 & 30-31, 2012
Kyushu University & University of Tokyo

Company	Subject
NTT Secure Platform Laboratories	Arithmetic-Cryptography
FUJITSU LABORATORIES LTD.	Integer factorization problems Arithmetic-Cryptography
Kao Corporation	Graph Theory-Optimization
OLM Digital, Inc.	CG of animation Geometry and Statistics
Nippon Steel Corporation	Multi-scale modeling, Anomalous diffusion Geometry-Topology-Probability-PDE
Railway Technical Research Institute	Space curves Differential Geometry
RIKEN / The University of Tokyo	Mathematical Physics



Institute of Mathematics for Industry (IMI)

First Industrial Math. Institute
in Japan, Founded in April 2011



Institute of Mathematics for Industry
Kyushu University

Organization of IMI (27 full-time staffs)

- ◎ Advanced Math. Technology Section
- ◎ Applied Math. Research Section
- ◎ **Fundamental Math. Section**
Researchers of pure math., interested in application
- ◎ Laboratory of Advanced Software in Mathematics
- ◎ Visitor Section (from Academia & Industry)

Nikkei Newspaper (June 2010)

九州大学は来年、その名も「マ
ス・フォア・インダストリ（産
業数学）研究所」という新しい
研究所を設ける。産業界と協力
して産業界で役立つ数学者の育
成を目指す。

中外時評
企業は数式を愛せるか
もと産業数学の研究を

産業数学研
九大創設へ
来春、アジア初
中韓豪と連合体も

Nishinippon Newspaper
(Feb. 2010)

R&D in
Companies

Carrier Pass for
PhD increased

University,
etc.

PhD numbers
increased
(new system for
degree)



















Joint research
with companies

Team
Research

Doctoral course (Master course)



















- Experiencing joint research
- Human exchange

Joint Projects (no joint projects before 2005)

ASAHI GLASS	FUJITSU  	NIPPON STEEL CORP.  
PANASONIC 	Nisshin Fire & Marine Insurance Co., Ltd.  	Mitsubishi Research Inst. Inc.
MAZDA MOTOR CORP.  	KDDI	HITACHI 
IBM Japan 	ETRI, Korea	OLM Digital, Inc. & WETA Digital, New Zealand (CREST) 
Studio Phones 	Nonprofit organization-- Science Accessibility Net	
NTT  	12R, Singapore	 Appeared from long-term internship  Progressing now
New Energy and Industrial Technology Development Organization	National Institute of Information and Communications Technology 	

Joint Projects (no joint projects before 2005)

I participate here !

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Journal of Math-for-Industry



Institute of Mathematics for Industry
Kyushu University

Journal of Math-for-Industry

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Vol.4

Vol.3

Vol.2

Vol.1

 Editorial Board

 Guide for Authors



The Journal of Math-for-Industry presents original research papers and survey papers with original view points in all scientific disciplines in which the mathematics or that in industry play a basic role. Articles by scientists in a variety of interdisciplinary areas are published.
Research Areas Include:

- 1 Significant applications of mathematics to industry, including feedback from industry to mathematics
- 2 New developments in Mathematics for industry
- 3 New developments in Mathematics

Access Full Text 

Editorial Board 

Guide for Authors 

What's new

[JMI2012B has been published! \(5 October 2012\)](#)

Institute of Mathematics for Industry & Faculty of Mathematics, Kyushu University

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We are waiting for your
submission to JMI!





II. Application of Singularity Theory to Visualization of Big Data

Osamu Saeki (IMI, Kyushu Univ.)

Joint work with **Shigeo Takahashi**
(Univ. of Tokyo)

November 28, 2012



§1. Visualization of Scalar Function Data

Level set

§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

M^n : differentiable manifold of dimension n (or a region in \mathbf{R}^n)

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which is called a **level set**.

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For $n = 2$, it is a curve; for $n = 3$, it is a surface, etc.

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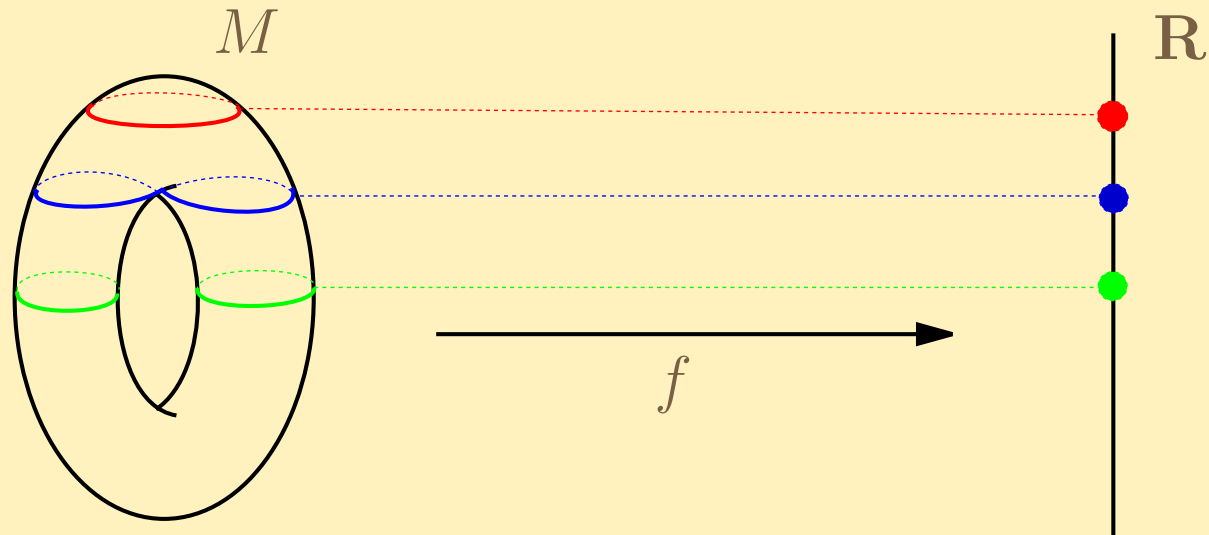
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Example 1.2 Altitude from the sea level (height function):
level set = contour line

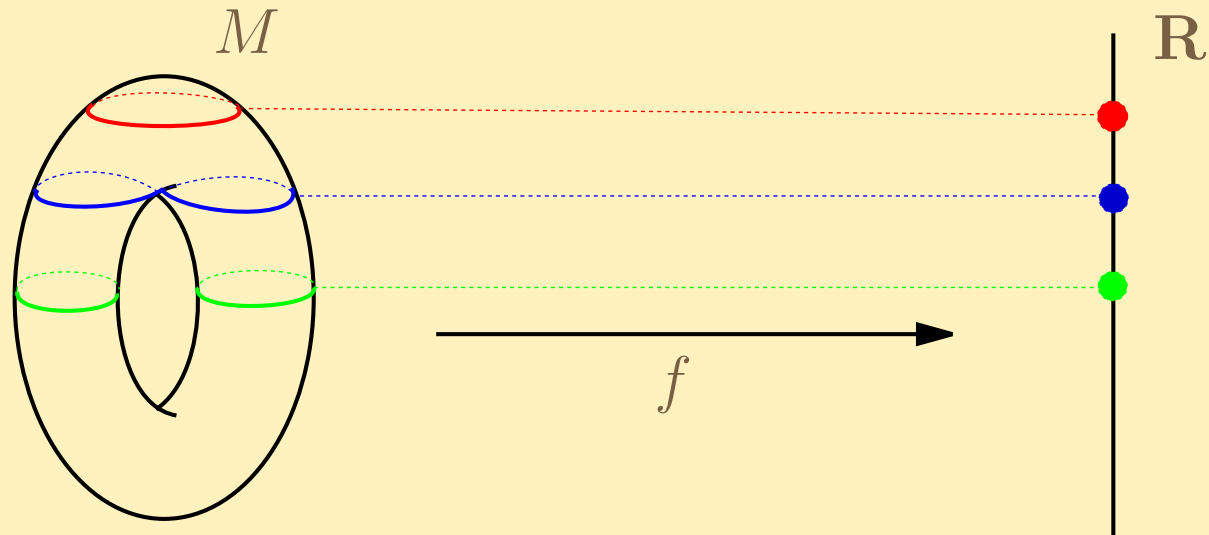
Example of level sets

§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data



Example of level sets

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A level set may not be connected.

Reeb graph

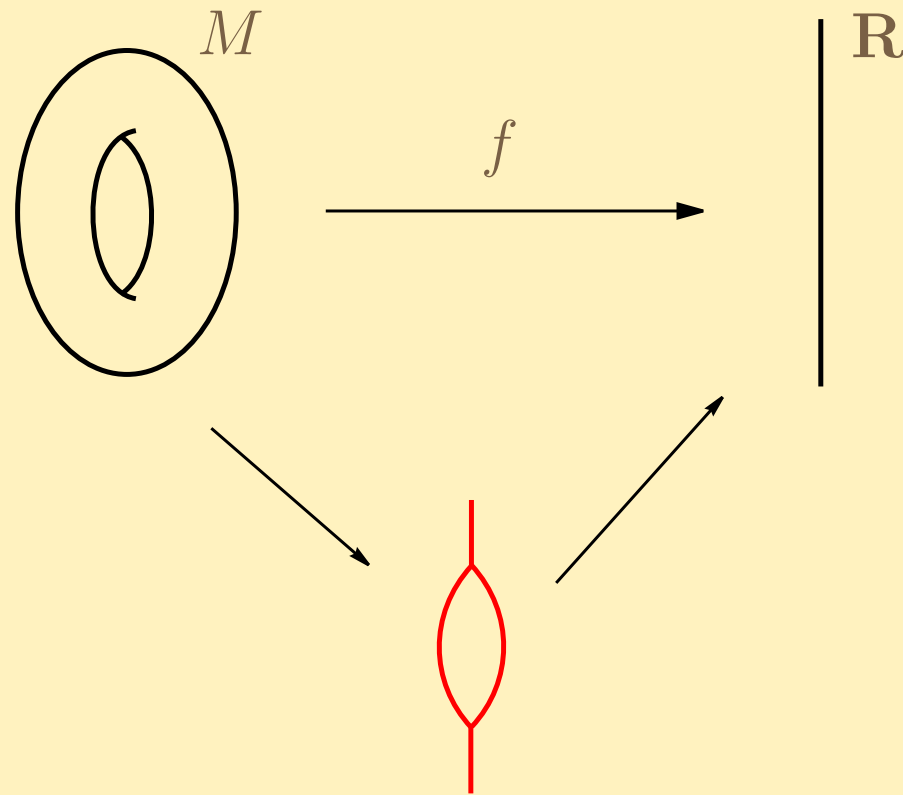
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The space (or graph) obtained by contracting each connected component of the level set to a point is called a **Reeb graph** (or contour tree, volume skeleton tree, Stein factorization, ...).

Reeb graph

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Reeb graph and visualization



§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

Reeb graph is very useful for visualizing 2 or 3-dimensional scalar functions.



Reeb graph and visualization



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Reeb graph and visualization

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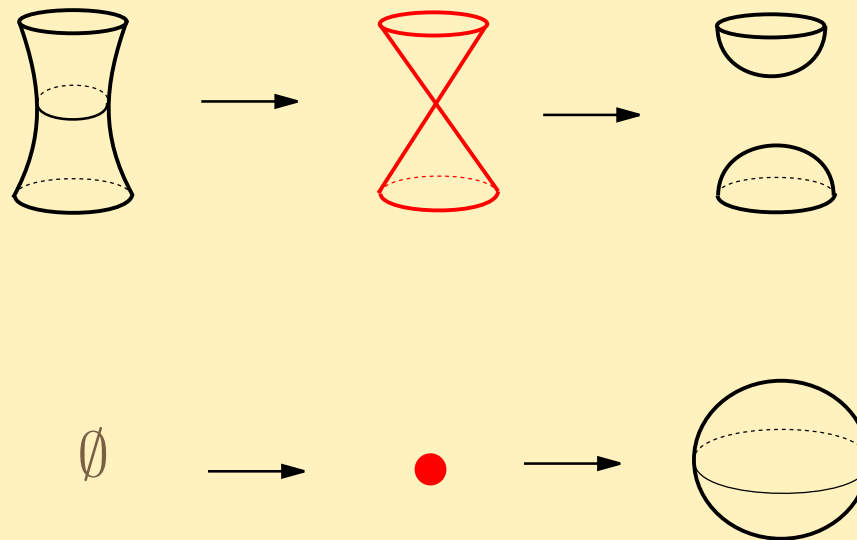
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- It is important to study the **topological change of the level sets** around each critical point.

Reeb graph and visualization

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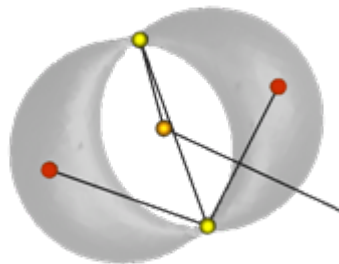
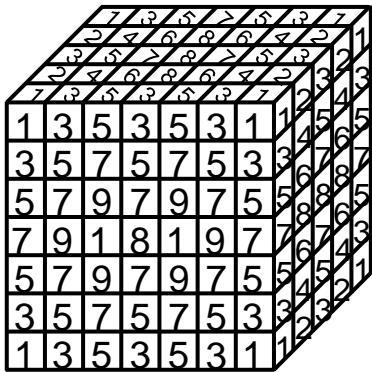
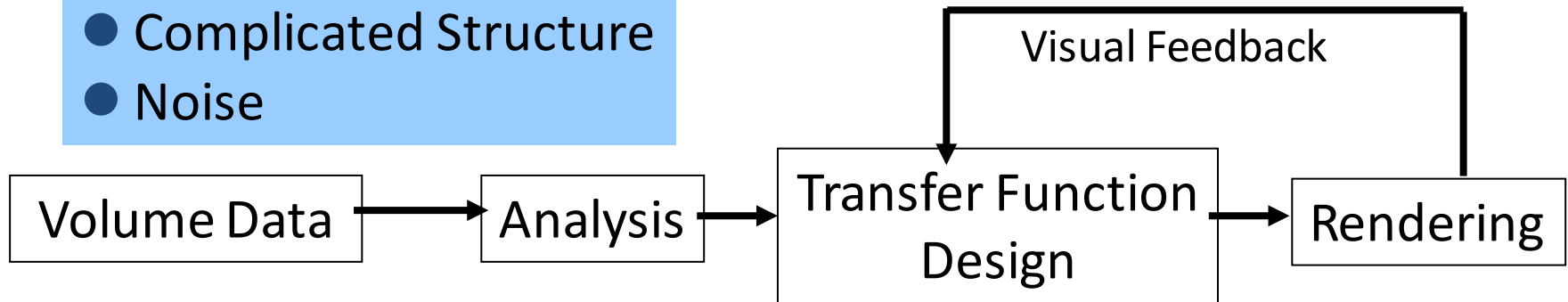
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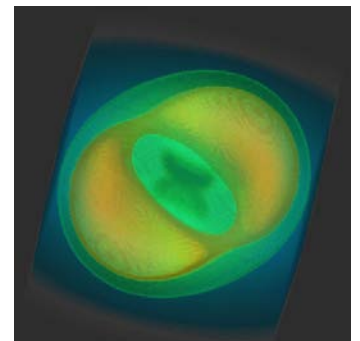
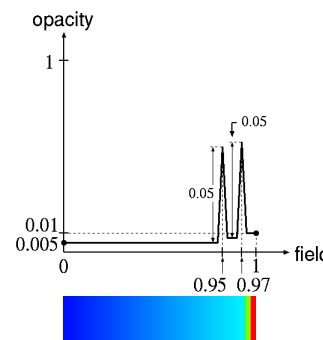
Example of level-surface change for a 3-dimensional scalar function

Direct Volume Rendering

- Big Size
- Complicated Structure
- Noise



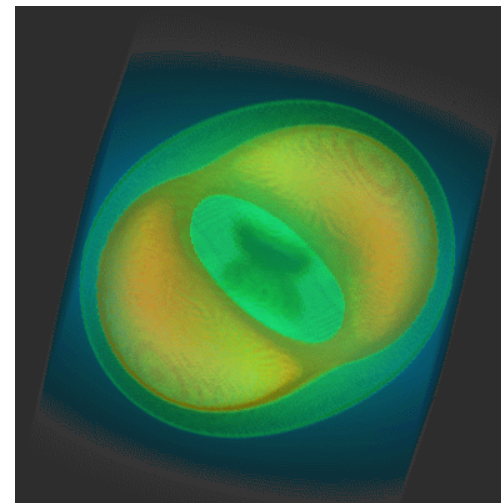
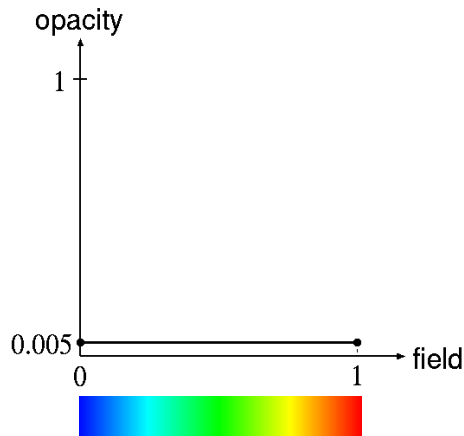
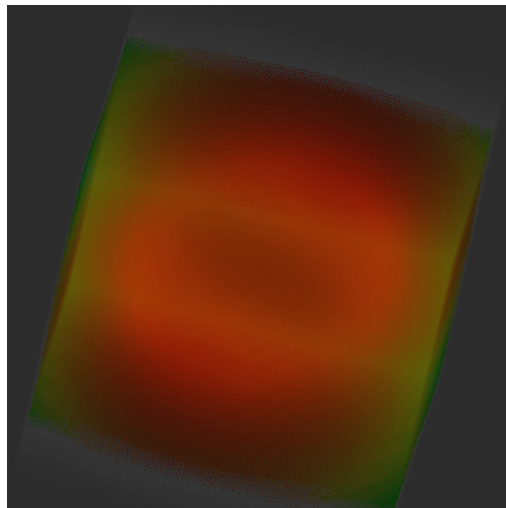
Reeb graph



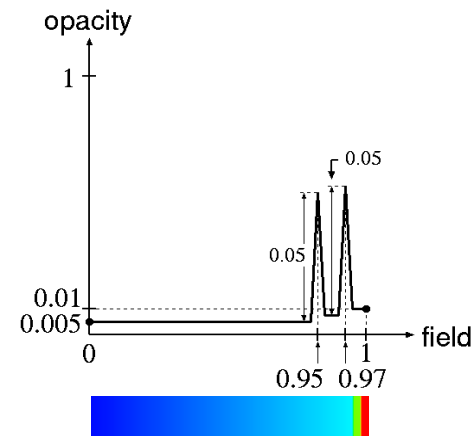
Example: an analytic function

Visualization Result (size 64^3)

Default TF

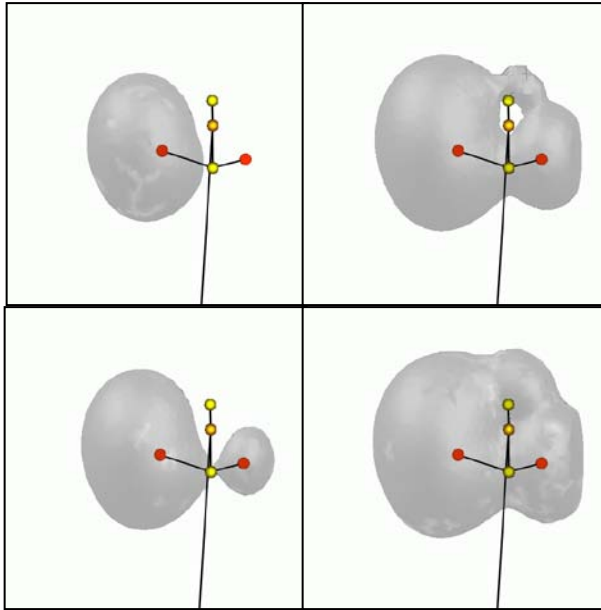


Accentuated
TF

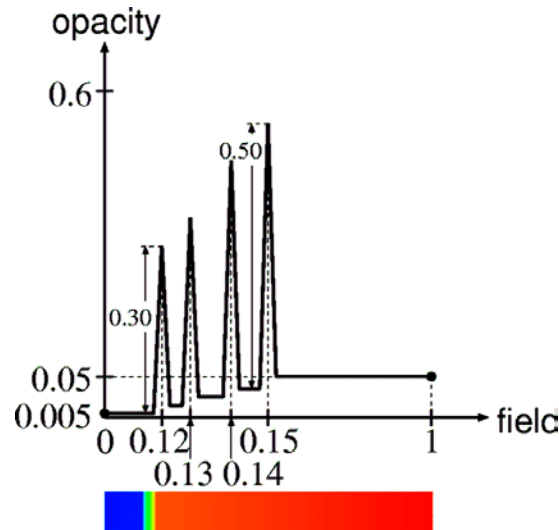


Example: Proton and Hydrogen-atom collision

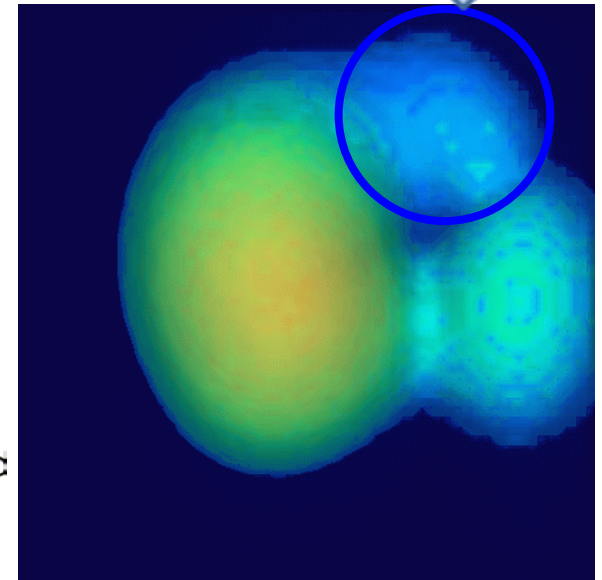
This was found by virtue of the topologically accentuated TF.



Characteristic
iso-surfaces



Designed TF



Visualization result

After the Collision

I am now engaged in a joint work with a **steel company** in Japan.

They can get **3D data** of steel materials by taking pictures of a lot of slices. (This is already not so easy!)



We can construct a Reeb graph (contour tree) to visualize the 3D data.



Can be used to estimate certain physical properties of the material (without doing any experiments that cost a lot).

In fact, certain technologies in **Topology** can be also useful!

Homology, Cohomology, Betti Numbers,
Euler Characteristics, Persistent Homology, ...



An Example of **MI**

§2. Visualization of Multi-function Data



Fiber



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Fiber

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$$f(x) = (f_1(x), f_2(x), \dots, f_m(x))$$

Fiber

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Definition 2.1 For $c \in \mathbf{R}^m$, $f^{-1}(c)$ is called a **fiber** (rather than a level set).

Example of fibers

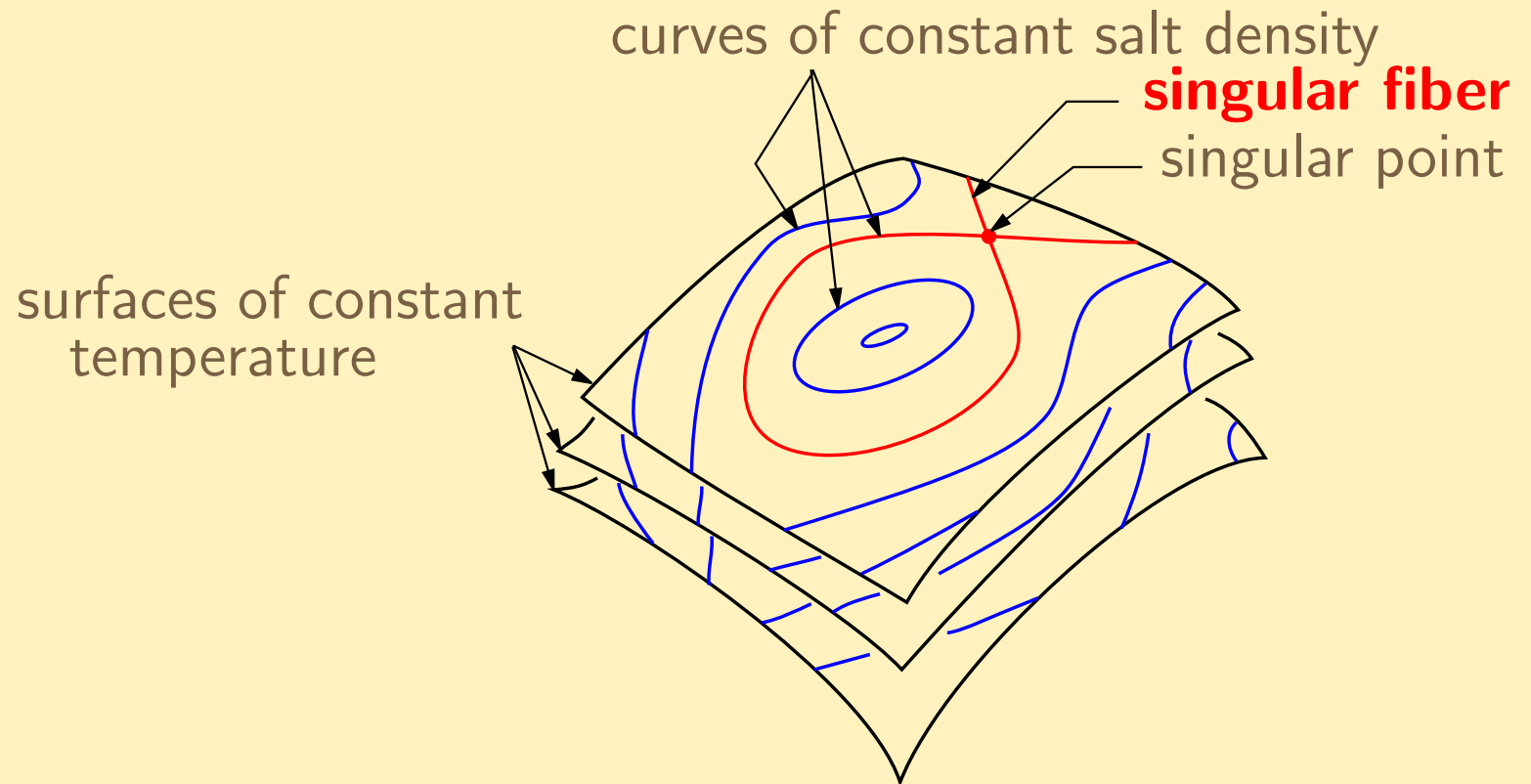
§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

$n = 3$, M^3 : sea water, $f : M^3 \rightarrow \mathbf{R}^2$
 $f = (\text{temperature}, \text{salt density})$

Example of fibers

§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

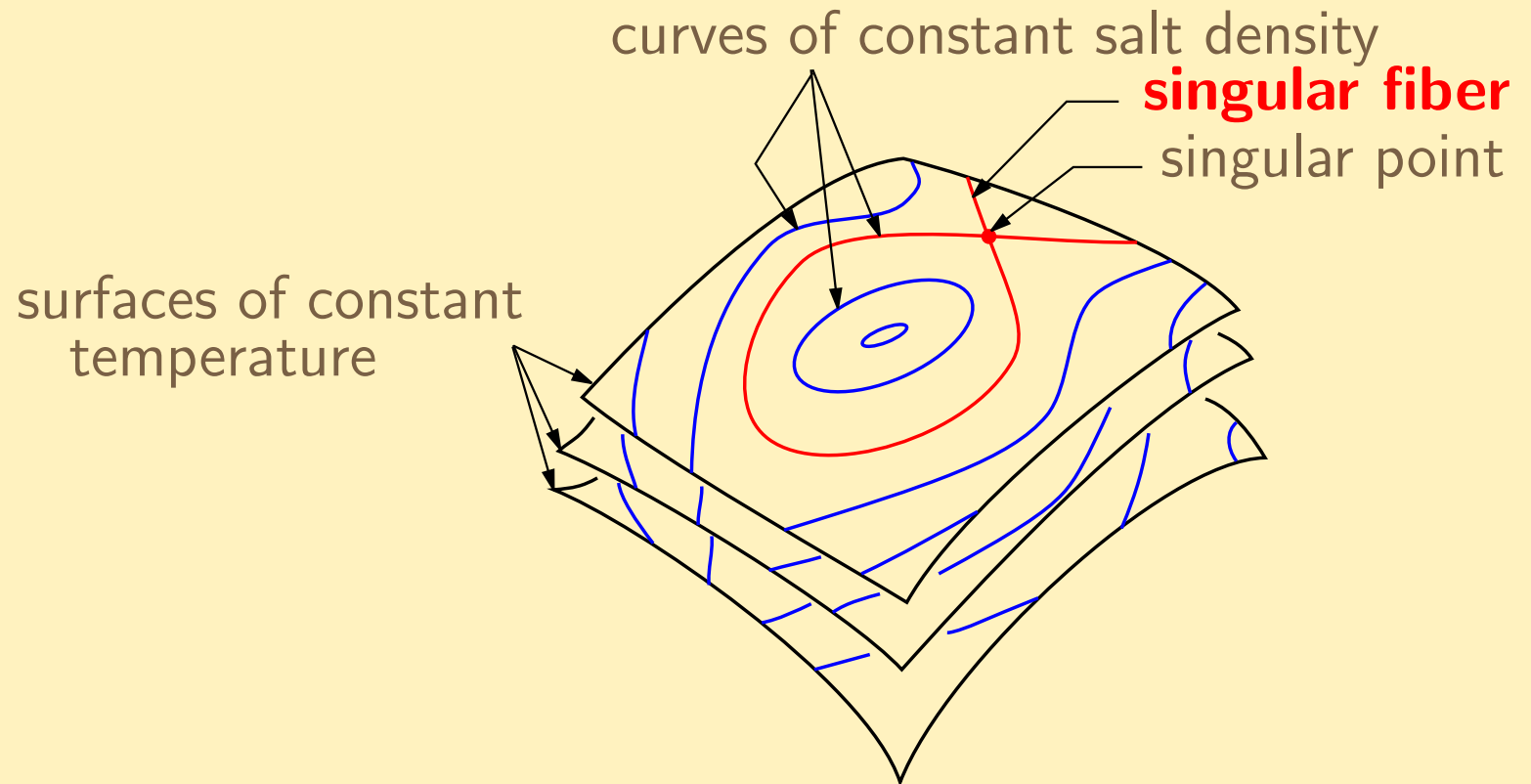
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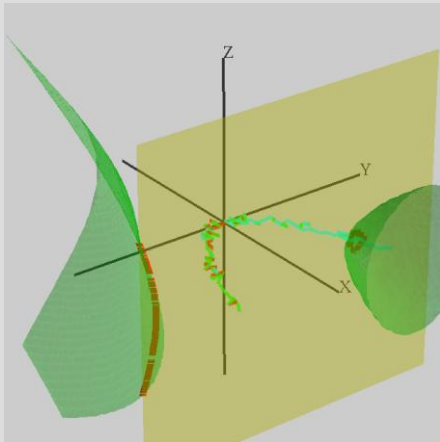


A fiber containing a singular point is called a **singular fiber**.



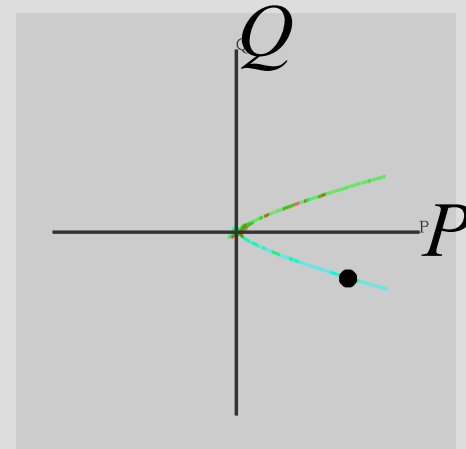
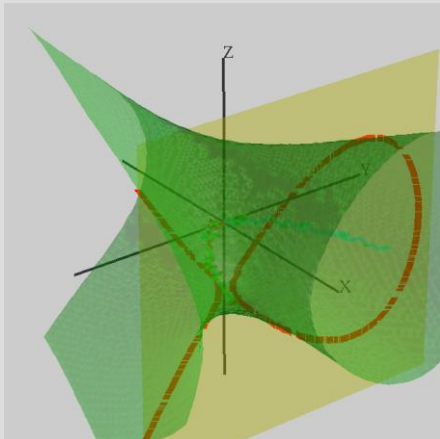
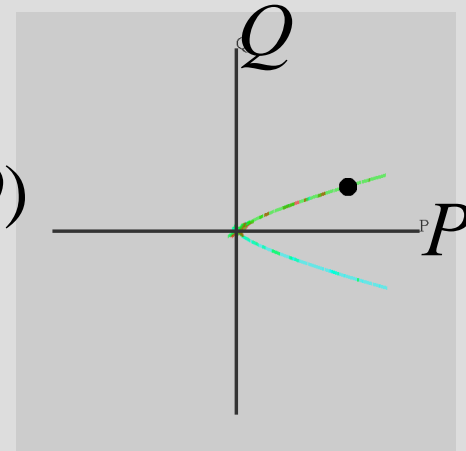
Singular Fibers

- Topology of the fibers change around singular fibers



$$f: \mathbf{R}^3 \rightarrow \mathbf{R}^2$$

$$(x, y, z) \rightarrow (P, Q)$$



Morse lemma

§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

$f : M^n \rightarrow \mathbb{R}$ differentiable function (scalar function)

Morse lemma

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critical point: a point in M^n where all 1st order partial derivatives of f vanish

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Theorem 2.2 (Morse lemma) *If f is **generic** enough, then around each critical point, f is expressed as*

$$f = \pm x_1^2 \pm x_2^2 \pm \cdots \pm x_n^2 + c$$

w.r.t. certain local coordinates for some constant c .

Morse lemma

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For the study of level-set changes, the Morse lemma is essential!

Morse lemma

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How about the case of multi-functions?



Singular points and Jacobi set



§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

$f : M^n \rightarrow \mathbf{R}^m$ ($n \geq m$) differentiable map (multi-function)

Singular points and Jacobi set

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$f : M^n \rightarrow \mathbf{R}^m$ ($n \geq m$) differentiable map (multi-function)

Definition 2.3 For a point $x \in M^n$,

$$df_x : T_x M^n \rightarrow T_{f(x)} \mathbf{R}^m$$

is the linear map associated with the **Jacobian matrix** of f (the $m \times n$ matrix whose entries are the first order partial derivatives).

Singular points and Jacobi set

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The set of singular points

$$J(f) = \{x \in M^n \mid \text{rank } df_x < m\}$$

is called the **Jacobi set** of f .

Singular points and Jacobi set

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In general, the Jacobi set $J(f)$ is of dimension $m - 1$.



Case of maps into \mathbb{R}^2



§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

Example 2.4 When $n = 2$ and $m = 2$.

Case of maps into \mathbb{R}^2

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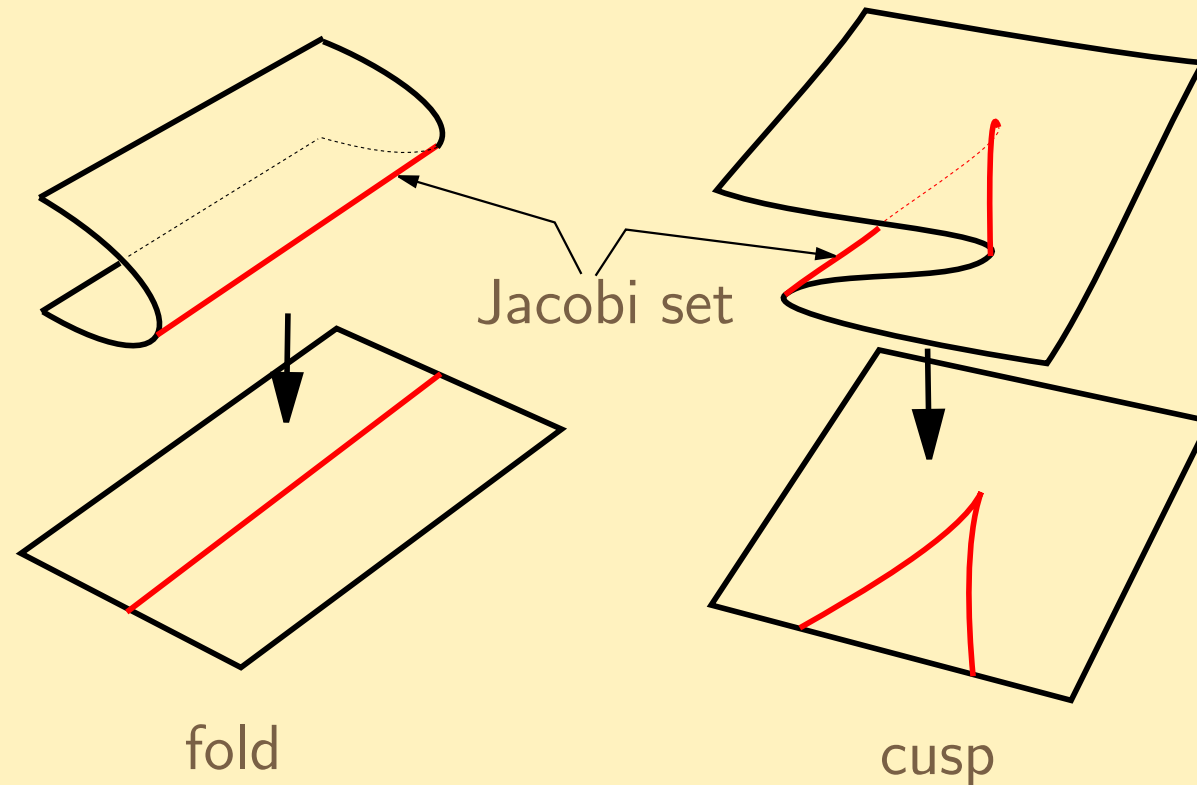
Types of singularities: **fold** and **cusp** (Whitney, 1955)

Case of maps into \mathbb{R}^2

§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

Example 2.4 When $n = 2$ and $m = 2$.

Types of singularities: **fold** and **cuspidal** (Whitney, 1955)



Singular points of multi-functions

§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

$$f : M^n \rightarrow \mathbf{R}^m$$

Suppose $n \geq m = 2$.

Singular points of multi-functions

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fold: A generalization of the Morse critical points for scalar functions

Singular points of multi-functions

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Suppose $n \geq m = 2$.

fold: A generalization of the Morse critical points for scalar functions

cusp: A degeneration of fold singularities

Singular points of multi-functions

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cusp: A degeneration of fold singularities

For $m = 3$, a **swallowtail** may appear, which is a degeneration of cusp singularities.

Singular points of multi-functions

§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

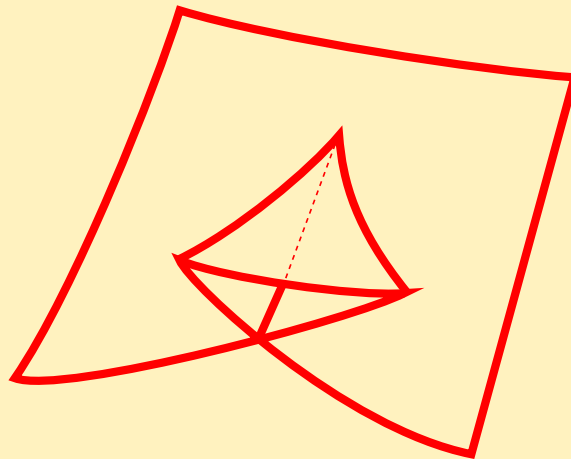
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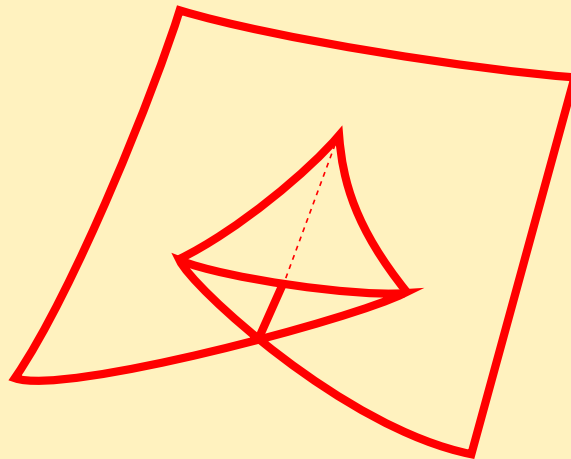
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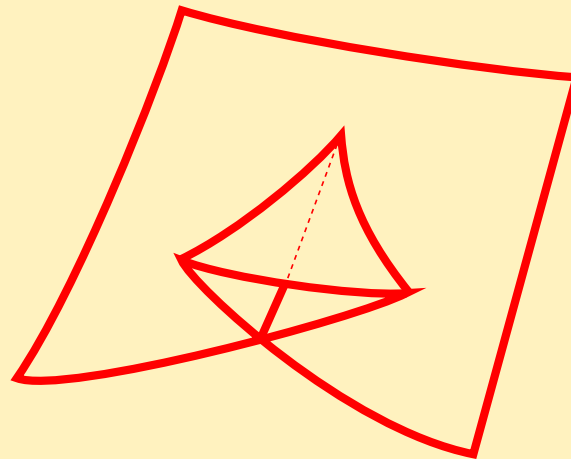
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\implies still extensively studied!

Identifying cusps

§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

$$f : M^n \rightarrow \mathbf{R}^m$$

Suppose $n - m > 0$ is odd.

Identifying cusps

§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

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$$\lambda = 0, 1, \dots, (n - m + 1)/2.$$

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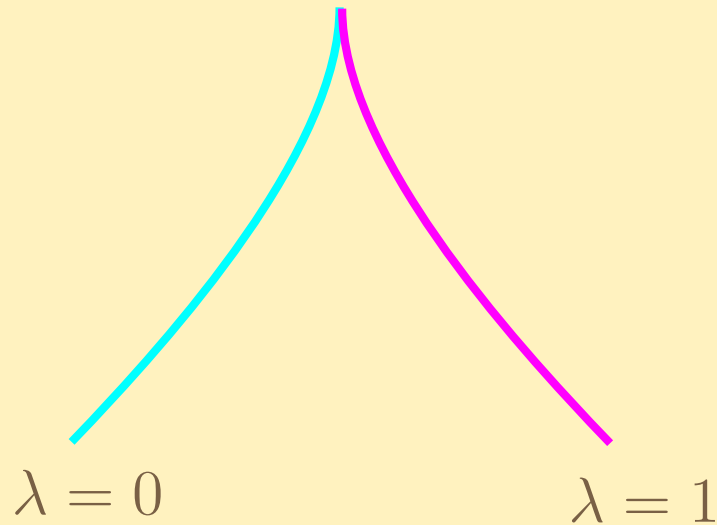
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When $n = 3$ and $m = 2$.



Algorithm



§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

Edelsbrunner–Harer (2002)

Suggested an algorithm for obtaining the Jacobi set.



Algorithm



§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

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Singularity theory of differentiable mappings



One can identify the singularity types (to a certain extent)

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Singularity theory of differentiable mappings



One can identify the singularity types (to a certain extent)

If one can identify the folds, cusps and swallowtails, this can contribute a lot to the **visualization** of big data sets.



For visualization



§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

For visualization of multi-function data, we need to



For visualization



§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

For visualization of multi-function data, we need to

1. Identify the Jacobi set



For visualization



§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

For visualization of multi-function data, we need to

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2. Identify the singularity types



For visualization



§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

For visualization of multi-function data, we need to

1. Identify the Jacobi set
2. Identify the singularity types
3. Identify the Jacobi set image



For visualization



§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

For visualization of multi-function data, we need to

1. Identify the Jacobi set
2. Identify the singularity types
3. Identify the Jacobi set image
4. Identify the (singular) fibers



For visualization



§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

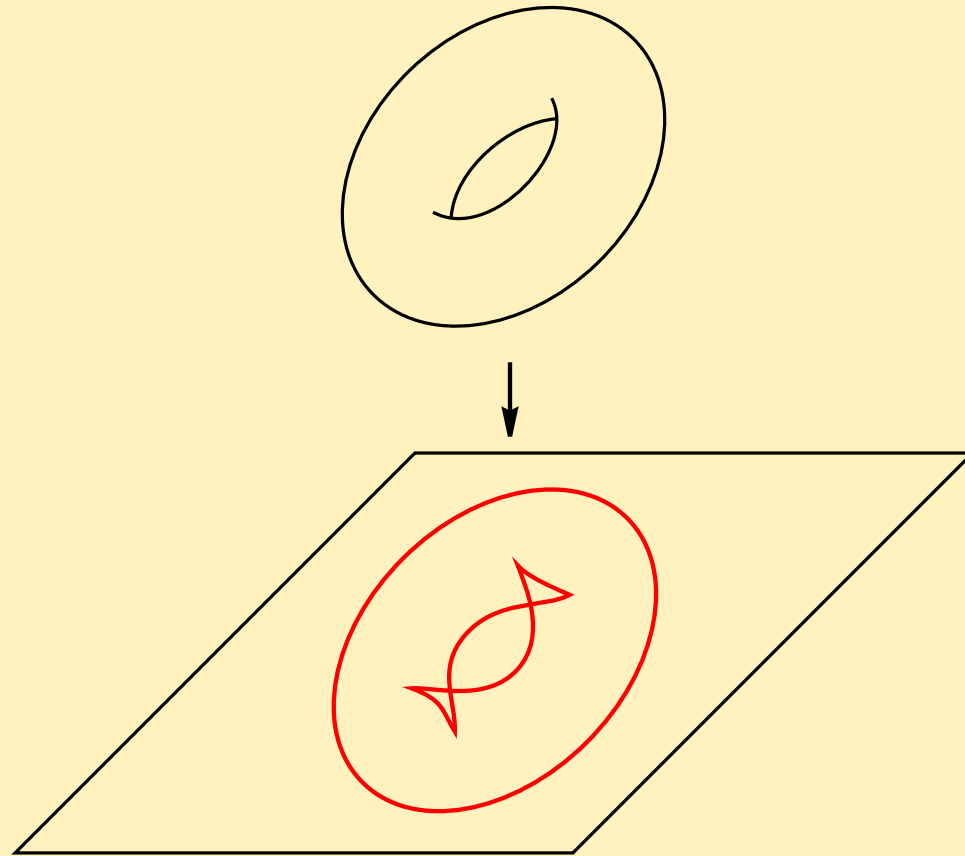
For visualization of multi-function data, we need to

1. Identify the Jacobi set
2. Identify the singularity types
3. Identify the Jacobi set image
4. Identify the (singular) fibers

In particular, for item 4 above, it is essential to identify the **singular fibers** and the **fiber changes** near singular fibers.

Example of a Jacobi set image

§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data



Jacobi set image of a map of a surface into \mathbb{R}^2

When $n = 3, m = 2$

§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

When $n = 3, m = 2$:

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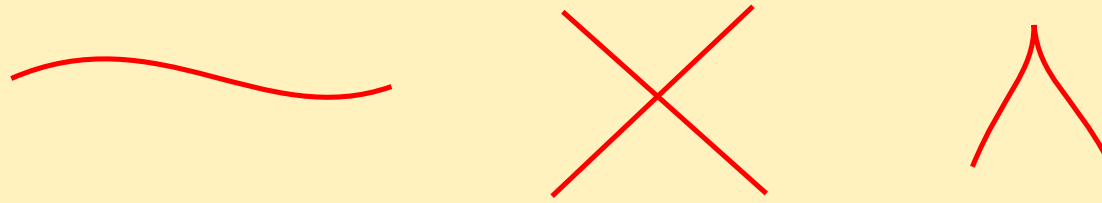
§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

When $n = 3, m = 2$: Jacobi set is a curve.

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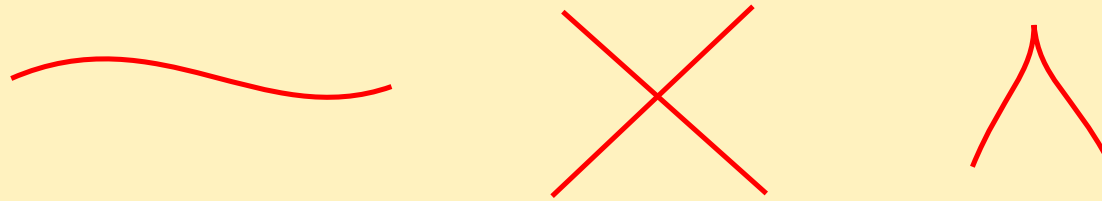


Local configurations of a Jacobi set image

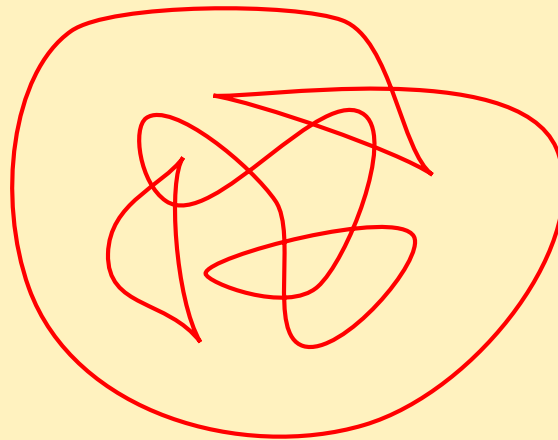
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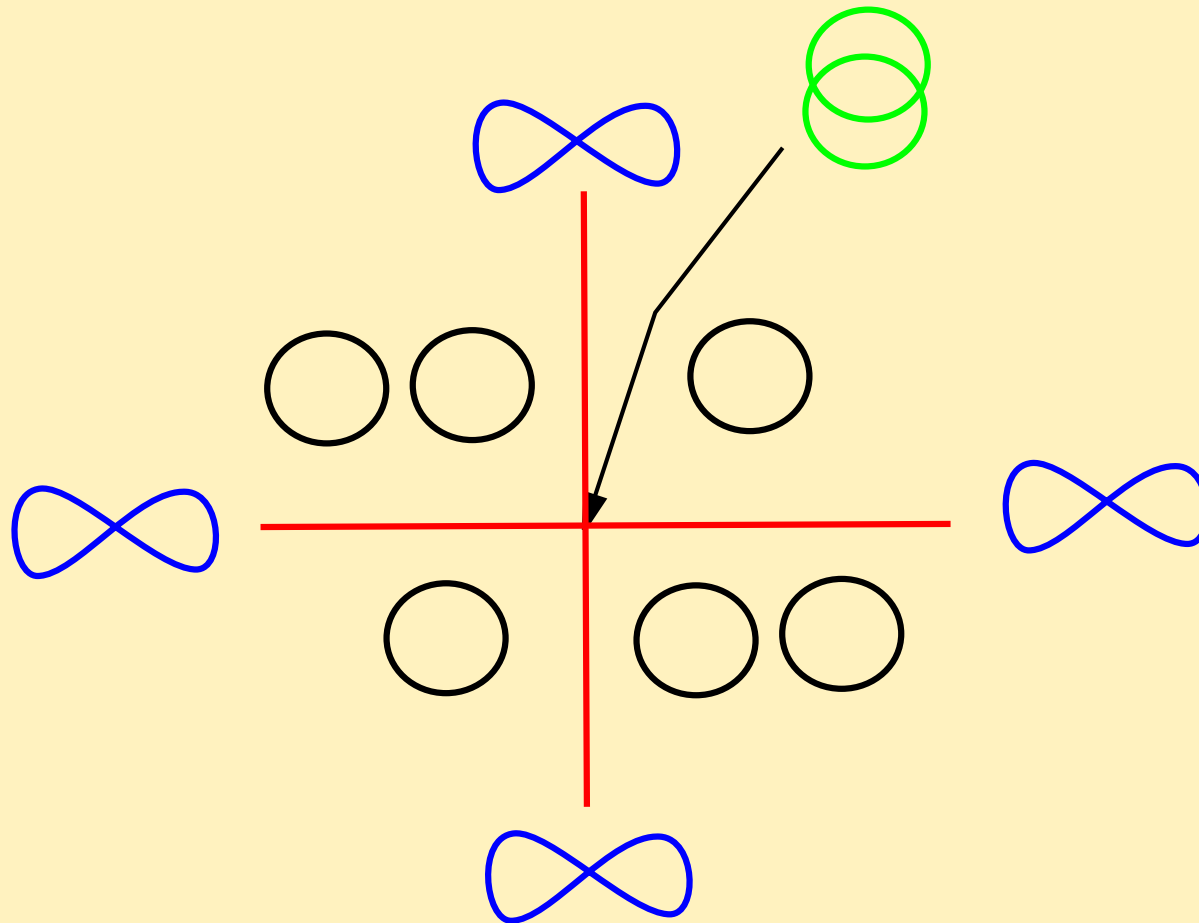
Local configurations of a Jacobi set image



Example of a Jacobi set image

Example of local fiber changes

§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data



Fiber over each region of $\mathbf{R}^2 \setminus f(J(f))$



Complexity of a fiber



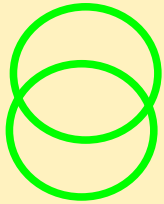
§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

Fibers are classified into some classes according to their complexities. This is measured by the complexity κ (called the **codimension**).

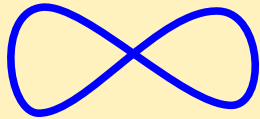
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§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

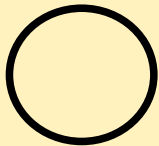
Fibers are classified into some classes according to their complexities. This is measured by the complexity κ (called the **codimension**).



Most complicated, $\kappa = 2$, appears discretely



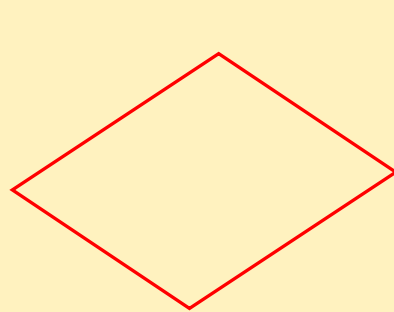
Moderately complicated, $\kappa = 1$, appears along curves



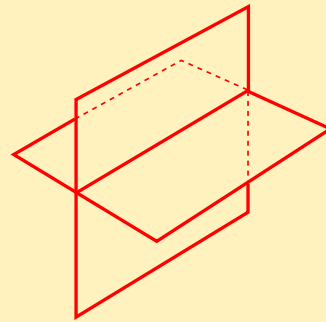
Most simple, $\kappa = 0$, appears along surfaces

When $n = 4$, $m = 3$

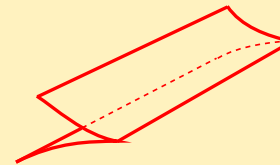
§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data



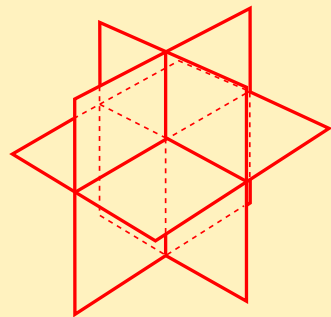
(1)



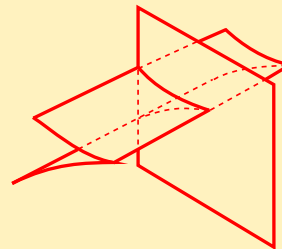
(2)



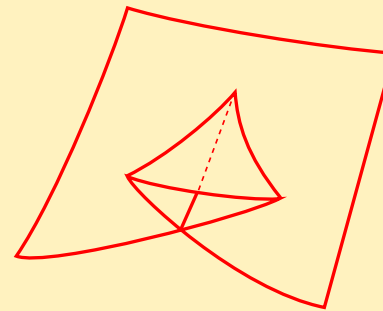
(3)



(4)



(5)

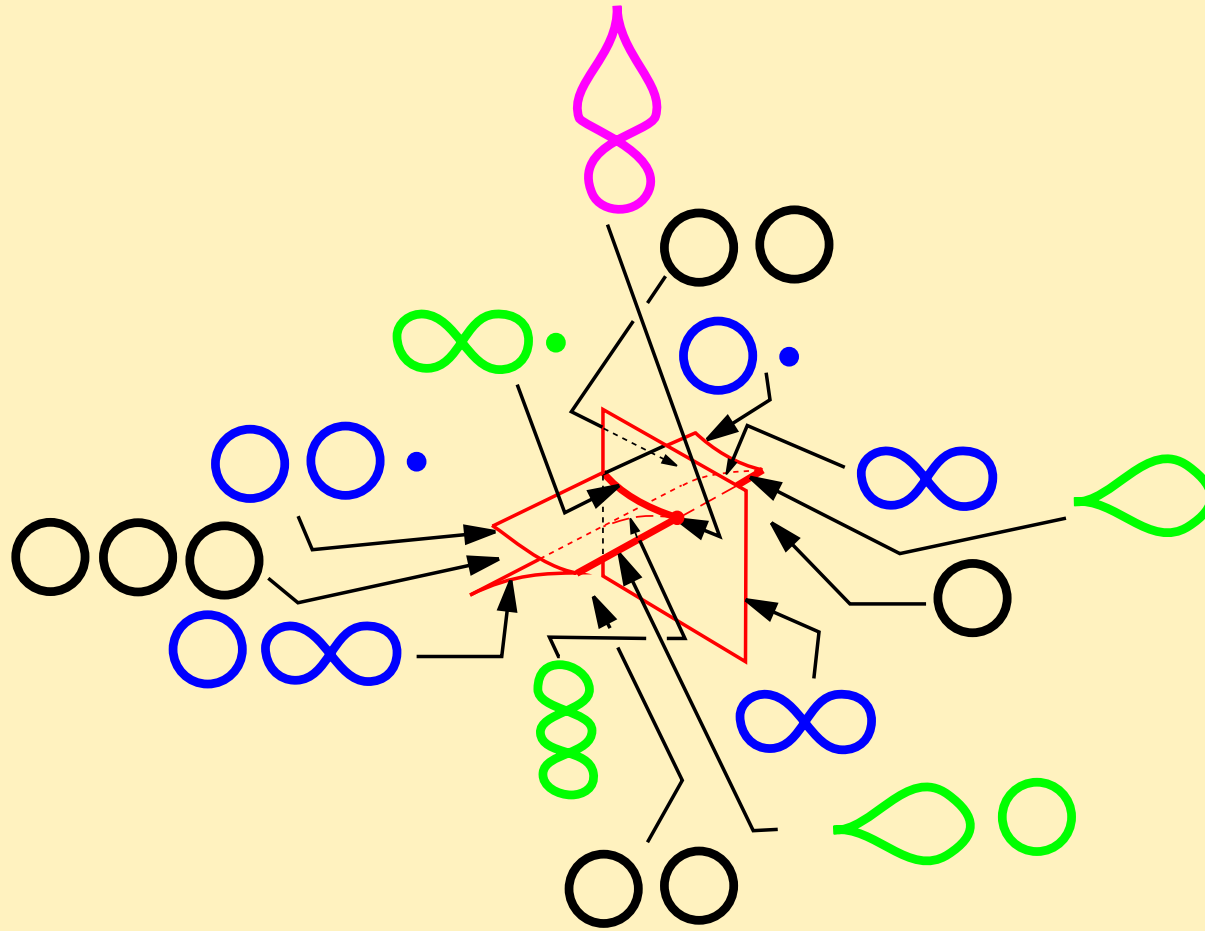


(6)

Local configurations of the Jacobi set image for maps $f : M^4 \rightarrow \mathbf{R}^3$

Example of fiber changes

§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data





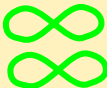
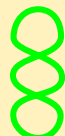
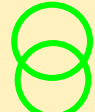
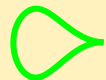


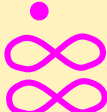
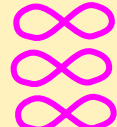
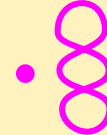
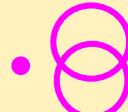
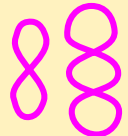
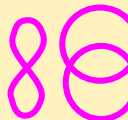

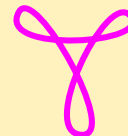
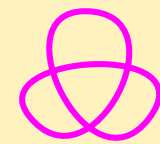
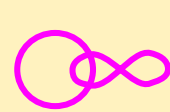


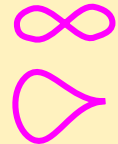

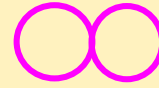
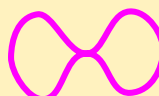



An example of fiber changes for a map $f: M^4 \rightarrow \mathbb{R}^3$

List of singular fibers

§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

When $n = 4, m = 3$

$\kappa = 1$						
$\kappa = 2$						
$\kappa = 3$						
						
						
						

Conclusion

§1. Visualization of Scalar Function Data §2. Visualization of Multi-function Data

By using the singularity theory of differentiable mappings,

- We can list up singularity types that appear generically.
- We can list up types of fibers.
- We can identify the singularities (or the singular fibers), and we can determine their types (to a certain extent).



still being investigated

**This contributes a lot to visualization
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Muito obrigado!