

# Conformal Geometry of the Visual Cortex

Matilde Marcolli and Doris Tsao

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Geometry of Neuroscience

## Functional Architecture of the V1 visual cortex

Filtering of optical signals by visual neurons and local differential data; integration of local differential data and global geometry, through global coherence of functional architecture of visual areas

This lecture is based on:

### References:

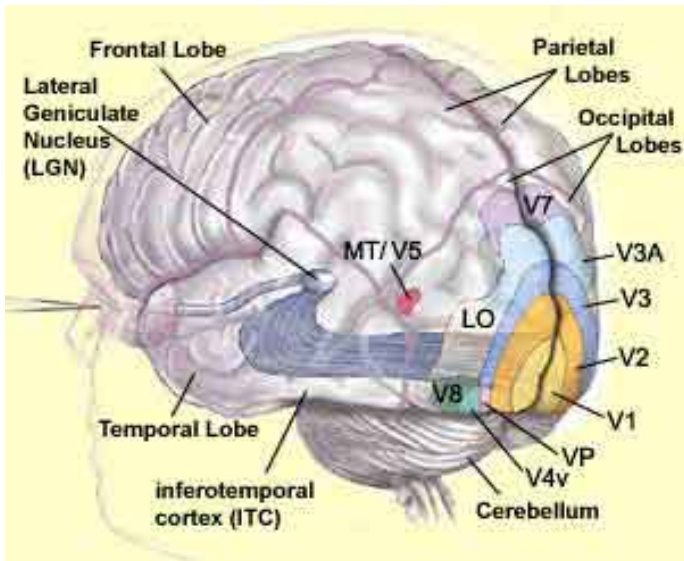
Pe Jeat Petitot, *Neurogéométrie de la vision*, Les Éditions de l'École Polytechnique, 2008

TSBBLW Duyan Taa, Jie Shia, Brian Bartonb, Alyssa Brewerb, Zhong-Lin Luc, Yalin Wang, *Characterizing human retinotopic mapping with conformal geometry: A preliminary study*, 2014

WGCTY Yalin Wang, Xianfeng Gu, Tony F. Chan, Paul M. Thompson, Shing-Tung Yau, *Intrinsic Brain Surface Conformal Mapping using a Variational Method*, Proceedings of SPIE Vol. 5370, 2004

## Areas of the visual cortex

- **V1**: the first of the primary visual areas (numerous feedbacks of successive areas like V2 and V4: here focus only on the geometry of V1)
- **high-resolution buffer hypothesis** of Lee–Mumford: V1 not just a bottom-up early-module but participating in all visual processes that require fine resolution
  - Lee, T.S., Mumford, D., Romero, R., Lamme, V.A.F., *The role of primary visual cortex in higher level vision*, Vision Research, 38 (1998) 2429–2454.



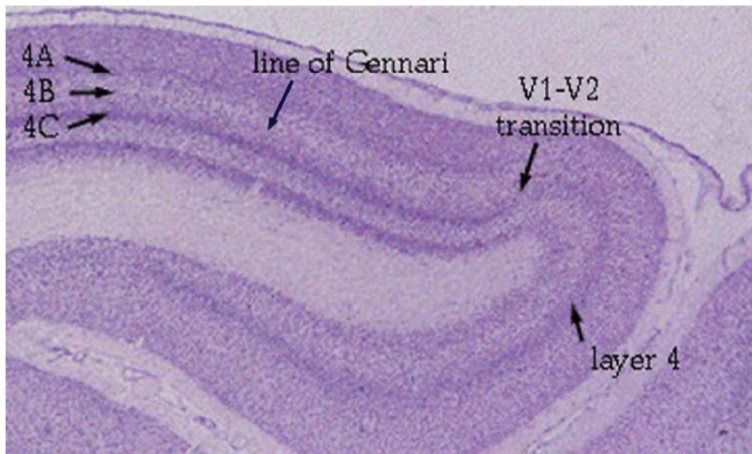
Location of the Visual Areas

## Structures in V1

- neurophysiology identifies three types of structures in primate V1
  - ① laminar
  - ② retinotopic (retinal mapping)
  - ③ (hyper)columnar

## Laminar Structure

- organized in 6 distinct horizontal layers (parallel to the surface of the cortex)
- look in particular at layer 4 (and sublayer 4C): main target of thalamocortical afferents and intra-hemispheric corticocortical afferents
- contains different types of stellate and pyramidal neurons

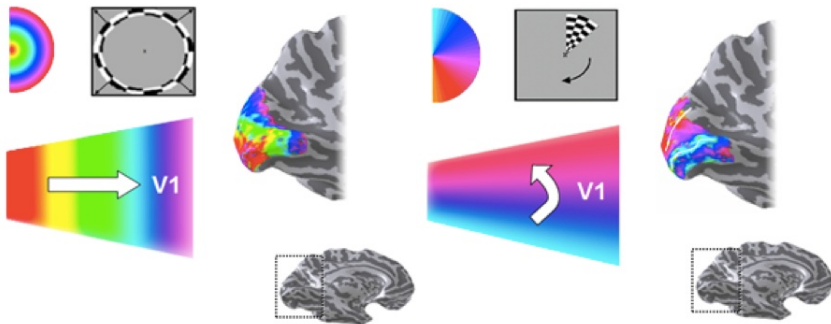


laminar structure and the 4th layer

## Retinotopy

- adjacent neurons with receptive fields covering overlapping portions of the visual field
- mapping of the visual input from the retina of the visual cortex are **conformal maps** (preserving local shape and local angles, but not distances and sizes)
- **logarithmic conformal mapping** from the retina to the sublayer 4C of layer 4 of the laminar structure
- Note: in cortical areas other than V1 adjacent points of the visual field may be mapped to non-adjacent regions

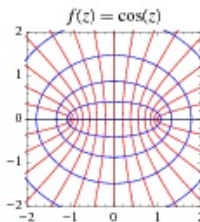
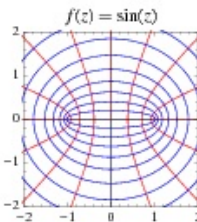
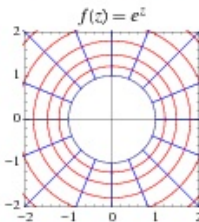
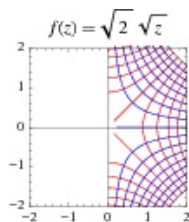
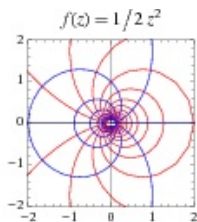
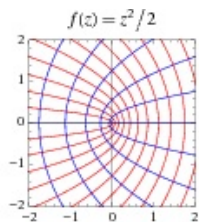


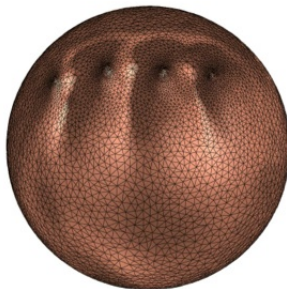


eccentricity and polar angle data (from [TSBBLW])

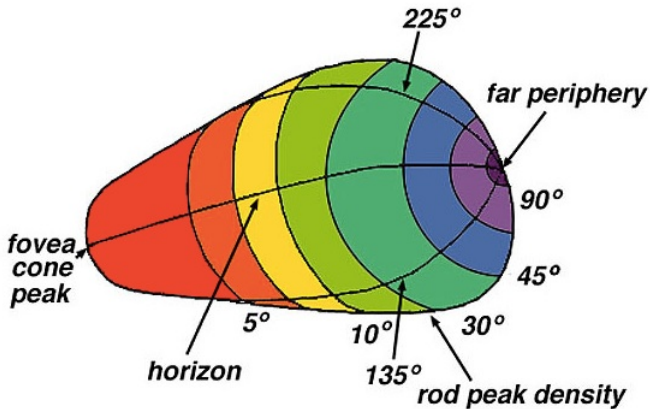
## Conformal maps

biholomorphic maps  $w = f(z)$  where  $f'(z) \neq 0$





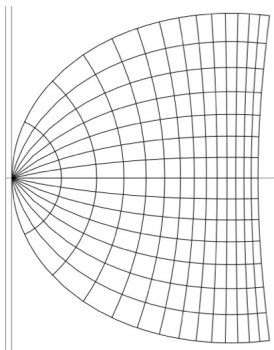
genus zero surface conformally mapped to  $S^2$  (from [WGCTY])



the unfolded striate cortex with the mapping of the visual field

## Models of retinotopy conformal mapping

- the  $\log(z + a)$  model (also referred to as “monopole model”)

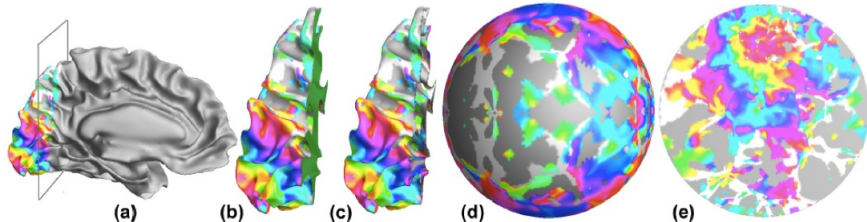


- more general  $\log\left(\frac{w(z)+a}{w(z)+b}\right)$  model (also known as “wedge-dipole model”)

TSBBLW Duyan Taa, Jie Shia, Brian Bartonb, Alyssa Brewerb,  
Zhong-Lin Luc, Yalin Wang, *Characterizing human retinotopic  
mapping with conformal geometry: A preliminary study*, 2014

- two step procedure to modeling retinotopy by conformal mapping
  - ① conformal map from brain visual cortex to the unit disk
  - ② conformal map from visual field to the unit disk

## From the visual cortex to the unit disk: conformal flattening



(1) slice along plane to isolate visual cortex regions; (b) visual regions after slicing; (c) double covering; (d) projection of double covering to a sphere; (e) stereographic projection to the unit disk (from [TSBBLW])

## Mesh and $u, v$ -coordinates

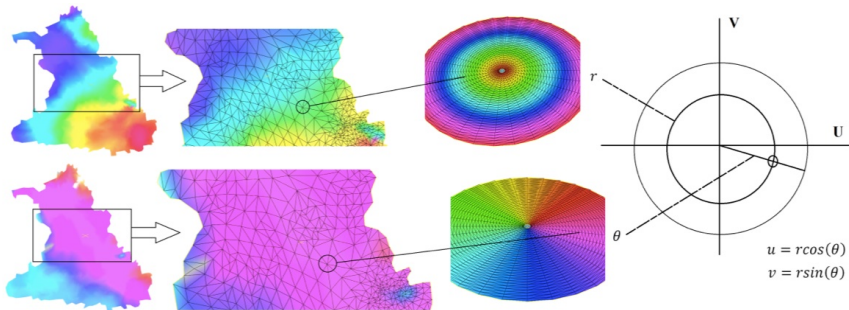
Data collection provides:

- simplicial complex (mesh triangulation)  $K$  of cortical area
- color gradient data for eccentricity and polar angle:  
parameterization of visual stimulus in the visual field as  
 $u = r \cos(\theta)$  and  $v = r \sin(\theta)$

general technique for constructing conformal mapping from

**WGCTY** Yalin Wang, Xianfeng Gu, Tony F. Chan, Paul M. Thompson, Shing-Tung Yau, *Intrinsic Brain Surface Conformal Mapping using a Variational Method*, Proceedings of SPIE Vol. 5370, 2004





mesh  $K$  with color gradient data for eccentricity and polar angle  
determining  $u, v$ -coordinates at each vertex of the mesh  
(from [TSBBLW])

Constructing the conformal maps: energy minimizing [WGCTY]

- piecewise linear functions  $\mathcal{C}^{PL}(K)$ , quadratic form

$$\langle f_1, f_2 \rangle = \frac{1}{2} \sum_{e \in E(K)} k_e (f_1(s(e)) - f_1(t(e))) (f_2(s(e)) - f_2(t(e)))$$

$e \in E(K)$  edges,  $s(e), t(e) \in V(K)$  source and target vertices;  
 $k_e > 0$  parameters

- Energy functional

$$E(f) = \langle f, f \rangle = \sum_e k_e \|f(s(e)) - f(t(e))\|^2$$

when all  $k_e = 1$ : Tutte energy

- discrete Laplacian

$$\Delta(f) = \sum_e k_e (f(t(e)) - f(s(e)))$$

energy minimizing  $f$  satisfies  $\Delta(f) = 0$

- for **vector valued** functions: apply  $\Delta$  componentwise
- $f : K_1 \rightarrow K_2$  map between two meshes (embedded in Euclidean spaces  $\mathbb{E}^3$ )

$$(\Delta f(v))^\perp = \langle \Delta f(v), \vec{n}(f(v)) \rangle \vec{n}(f(v))$$

**normal component**, with  $\vec{n}(f(v))$  normal vector to  $K_2$  at  $f(v)$

- **harmonic map**  $f : K_1 \rightarrow K_2$  iff  $\Delta f(v) = (\Delta f(v))^\perp$  (only normal no tangential component)
- vanishing of **absolute derivative**

$$Df(v) = \Delta f(v) - (\Delta f(v))^\perp$$

conformal maps to  $S^2$  by steepest descent [WGCTY]

- **non-uniqueness** of solutions: action of Möbius transformations on  $S^2 = \mathbb{P}^1(\mathbb{C})$

$$\mathrm{GL}_2(\mathbb{C}) \ni \gamma = \begin{pmatrix} a & b \\ c & d \end{pmatrix} : z \mapsto \frac{az + b}{cz + d}$$

- **constraints** to obtain a unique solutions:
  - **zero-mass** constraint:  $f : K_1 \rightarrow K_2$

$$\int f d\sigma_{K_1} = 0$$

- **landmark** constraints: manually labelled set of curves or point sets, optimal Möbius transformation that reduces distance between images of landmarks in the sphere  $S^2$

**Algorithm 1** [WGCTY] (steepest descent with Tutte energy)

- 1 start with mesh  $K$  and Gauss map  $\tau : K \rightarrow S^2$  with  $N(v) = n(v)$  normal to  $K \subset \mathbb{E}^3$
- 2 compute Tutte energy  $E_0 = E(\tau)$
- 3 compute absolute derivative  $D\tau(v)$
- 4 update  $\tau$  by  $\delta\tau = -D\tau(v) \cdot \delta t$  (fixed increment length  $\delta t$ )
- 5 compute Tutte energy: if  $E_{new} < E_0 + \delta E$  (fixed threshold  $\delta E$ ) output, else update  $E_0$  to  $E$  and repeat

Unique minimum, convergence to Tutte embedding of graph (1-skeleton of  $K$ ) in the sphere  $S^2$

## Algorithm 2 [WGCTY] (from Tutte embedding to conformal map)

- 1 compute Tutte embedding  $\tau$  as before and its Tutte energy  $E_0$
- 2 compute absolute derivative  $D\tau(v)$  and update  $\delta\tau(v) = -D\tau(v)\delta t$
- 3 compute Möbius transformation  $\gamma_0 : S^2 \rightarrow S^2$  that minimizes norm of the mass center

$$\gamma_0 = \operatorname{argmin}_{\gamma} \left\| \int \gamma \circ \tau d\sigma_K \right\|^2$$

- 4 compute harmonic energy: where coefficients  $k_e = a_e^\alpha + a_e^\beta$  (for edge  $e$  in boundary of faces  $F_\alpha$  and  $F_\beta$ )

$$a_e^\alpha = \frac{1}{2} \frac{(s(e) - v) \cdot (t(e) - v)}{|(s(e) - v) \times (t(e) - v)|}$$

where  $v$  third vertex in triangle face  $F_\alpha$

- 5 if  $E < E_0 + \delta E$  output current function; otherwise update  $E_0$  to  $E$  and repeat

- used minimization of mass center norm by Möbius transformations, but also want to evaluate how good conformal parameterization is, with respect to some given landmarks
- suppose obtained two parameterizations  $f_i : S^2 \rightarrow S$ , compare them in terms of given landmarks
- formulate again in terms of an energy functional

$$E(f_1, f_2) = \int_{S^2} \|f_1(u, v) - f_2(u, v)\|^2 du dv$$

look for Möbius transformation  $\gamma_\star$  that minimizes this energy

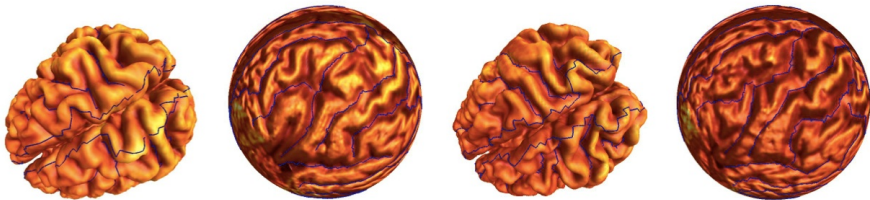
$$\gamma_\star = \operatorname{argmin}_\gamma E(f_1, f_2 \circ \gamma)$$

- using landmarks to only compare over a finite set of points (or over some assigned curves)
- say landmarks are finite sets of points  $\mathcal{P} \subset S_1$  and  $\mathcal{Q} \subset S_2$  with bijection  $p_i \leftrightarrow q_i$ ,  $i = 1, \dots, n$  between their preimages on  $S^2$
- look for Möbius transformation  $\gamma$  that minimizes

$$E(\gamma) = \sum_{i=1}^n \|p_i - \gamma(q_i)\|^2$$

non-linear problem, but assuming  $\gamma(\infty) = \infty$  by stereographic projection transform into a least square problem





landmark constraints: matching along preassigned curves, minimize  
landmark mismatch for representations from different subjects  
(from [WGCTY])

## Spherical harmonics orthonormal basis for $L^2(S^2)$

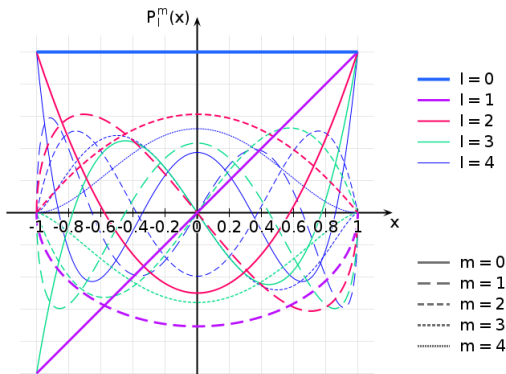
- $\ell \in \mathbb{N}$  and  $m \in \mathbb{Z}$  with  $|m| \leq \ell$  (degree and order)

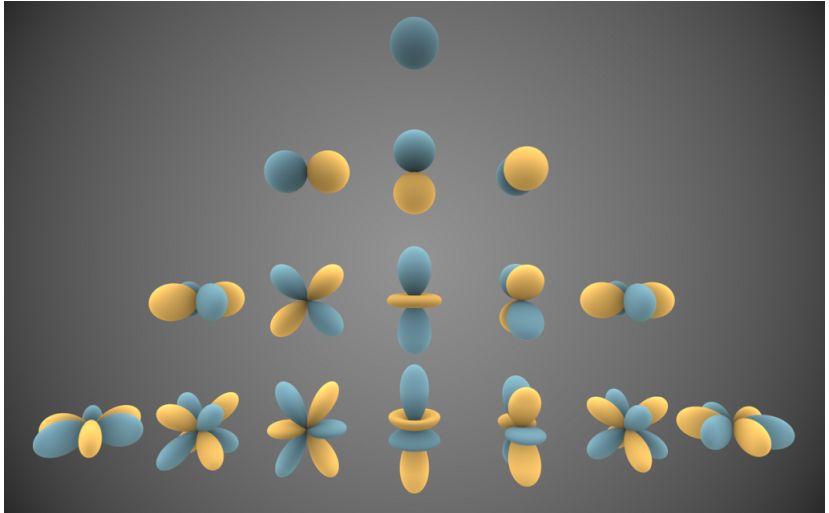
$$Y_{\ell}^m(\theta, \phi) = k_{\ell, m} P_{\ell}^m(\cos(\theta)) e^{im\phi}$$

$P_{\ell}^m$  associated Legendre polynomials

$$\frac{d}{dx}((1-x^2)\frac{d}{dx}P_{\ell}^m(x)) + (\ell(\ell+1) - \frac{m^2}{1-x^2})P_{\ell}^m = 0$$

associated legendre functions (normalized)





Real spherical harmonics  $\ell = 0, \dots, 3$ , yellow=negative, blue=positive, distance from origin=value in angular direction

- Expansion in spherical harmonics  $f \in L^2(S^2)$

$$f = \sum_{\ell \geq 0} \sum_{m: |m| \leq \ell} \langle f, Y_\ell^m \rangle Y_\ell^m$$

- suppose constructed conformal mapping of visual cortex to  $S^2$ , have coordinates on the cortex surface (embedded in  $\mathbb{E}^3$ )

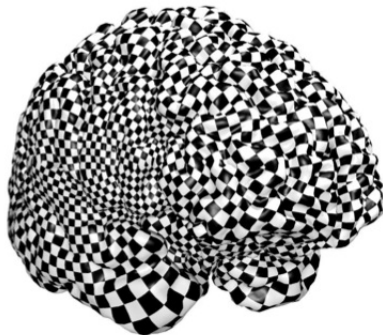
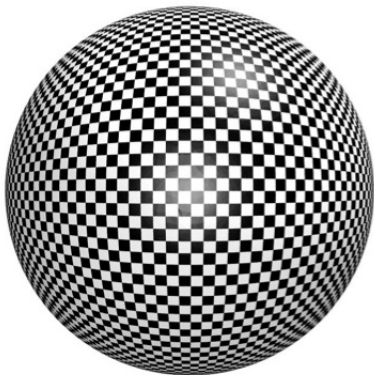
$$x^0(\theta, \phi), \quad x^1(\theta, \phi), \quad x^2(\theta, \phi)$$

with  $(\theta, \phi)$  angle coordinates on  $S^2$

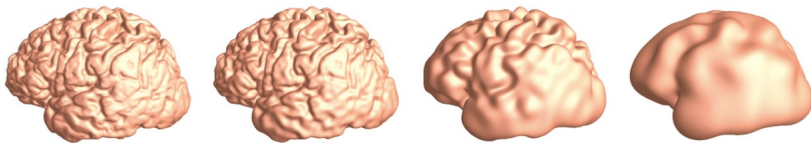
$$x^i(\theta, \phi) \in L^2(S^2), \quad \text{with} \quad \hat{x}^i(\ell, m) = \langle x^i, Y_\ell^m \rangle$$

coefficients of expansion in harmonic forms

- *Fast Spherical Harmonic Transform* to compute  $\hat{x}^i(\ell, m)$
- compression, denoising, feature detection, shape analysis: more efficiently performed on the Fourier modes  $\hat{x}^i(\ell, m)$

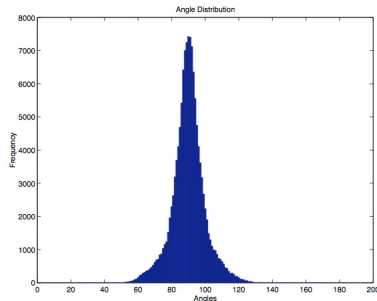
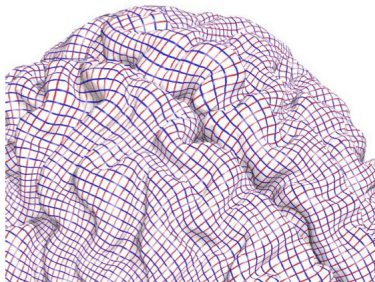


a conformal map from  $S^2$  to the brain surface (from [WGCTY])



geometric compression using low spherical harmonics and rescaling to smaller low frequencies coefficients (from [WGCTY])

## How good is modeling by conformal maps?



measuring deviation from conformality by deviation from right angle through inverse mapping from  $S^2$  to cortex surface  
(from [WGCTY])

## Beltrami equation and Beltrami coefficient

- a **conformal structure** at a point  $z \in \mathbb{C}$  is determined by a **complex dilatation**  $\mu(z)$  with  $|\mu(z)| < 1$
- intuitively, a conformal structure picks an ellipse centered at the origin as the new circle
- notation: for  $z = x + iy$

$$\frac{\partial}{\partial \bar{z}} = \frac{1}{2} \left( \frac{\partial h}{\partial x} + i \frac{\partial h}{\partial y} \right) \quad \text{and} \quad \frac{\partial}{\partial z} = \frac{1}{2} \left( \frac{\partial h}{\partial x} - i \frac{\partial h}{\partial y} \right)$$

- if  $\mu(z) = \mu$  constant, the function  $h(z) = z + \mu \bar{z}$  satisfies **Beltrami equation**

$$\frac{\partial h}{\partial \bar{z}} = \mu(z) \frac{\partial h}{\partial z}$$

- for constant  $\mu(z) = \mu$  round circle in  $h$ -plane corresponds to ellipse with constant  $|z + \mu \bar{z}|$  in  $z$ -plane: direction of axes from argument of  $\mu$  eccentricity from  $|\mu|$



- for  $\mu(z)$  real analytic: **Gauss isothermal coordinates**  $\exists$  local solution  $h(z)$  to Beltrami equation; Morrey for measurable  $\mu(z)$
- a solution  $h(z)$  on a local open set  $U$  is a **quasi-conformal mapping** with complex dilatation  $\mu(z)$
- conformal structure on a Riemann surface  $S$ : section of a disk  $D$  bundle over  $S$

$$\mu_\beta(z_\beta) = \mu_\alpha(z_\alpha) \frac{\partial z_\beta / \partial z_\alpha}{\partial \bar{z}_\beta / \partial \bar{z}_\alpha}$$

gluing of local  $\mu_\alpha : U_\alpha \rightarrow D$  on overlaps

- **Beltrami differential** on a Riemann surface  $S$  is antilinear homomorphism of tangent spaces  $T_z S$
- local solutions  $h_\alpha$  of Beltrami equation determine conformal coordinates for a Riemann surface  $S_\mu$  topologically equivalent to  $S$  but with a new conformal structure.

- in **genus zero** case: by Uniformization Theorem  $S_\mu$  is conformally equivalent to  $S = \mathbb{P}^1(\mathbb{C})$  with unique conformal isomorphism  $h$  that fixes  $\{0, 1, \infty\}$
- view  $h$  as **quasiconformal isomorphism** with dilatation  $\mu(z)$

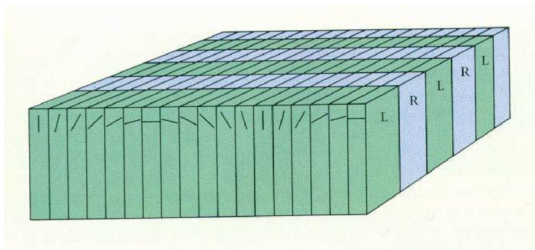
$$h : \mathbb{P}^1(\mathbb{C}) \rightarrow \mathbb{P}^1(\mathbb{C})$$

- conclusion from [TSBBLW]: compute Beltrami coefficient  $\mu$  for regions of V1 and V2 where reasonably smooth eccentricity and polar angle data: conformal map is very good approximation
- from the neuroscience point of view: **why** conformal maps?

## Columnar Structure

- another type of geometric structure present in visual cortex V1
- Hubel–Wiesel: columnar structures in V1: neurons sensitive to orientation record data  $(z, \ell)$ 
  - $z$  = a position on the retina
  - $\ell$  = a line in the plane
- local product structure

$$\pi : \mathcal{R} \times \mathbb{P}^1 \rightarrow \mathcal{R}$$



## Fiber bundles

- topological space (or smooth differentiable manifold)  $E$  with base  $B$  and fiber  $F$  with
  - surjection  $\pi : E \twoheadrightarrow B$
  - fibers  $E_x = \pi^{-1}(x) \simeq F$  for all  $x \in B$
  - open covering  $\mathcal{U} = \{U_\alpha\}$  of  $B$  such that  $\pi^{-1}(U_\alpha) \simeq U_\alpha \times F$  with  $\pi$  restricted to  $\pi^{-1}(U_\alpha)$  projection  $(x, s) \mapsto x$  on  $U_\alpha \times F$
- **sections**  $s : B \rightarrow E$  with  $\pi \circ s = id$ ; locally on  $U_\alpha$

$$s|_{U_\alpha}(x) = (x, s_\alpha(x)), \quad \text{with } s_\alpha : U_\alpha \rightarrow F$$

- **model of V1**: bundle  $\mathcal{E}$  with base  $\mathcal{R}$  the retinal surface, fiber  $\mathbb{P}^1$  the set of lines in the plane
- topologically  $\mathbb{P}^1(\mathbb{R}) = S^1$  (circle) so locally V1 product  $\mathbb{R}^2 \times S^1$
- We will see this leads to a geometric models of V1 based on

## Contact Geometry