

## Beltrami Holomorphic Flow (Bojarski)

Theorem: (Beltrami Holomorphic flow on  $\mathbb{S}^2$ ) There is a one-to-one correspondence between the set of quasiconformal diffeomorphisms of  $\mathbb{S}^2$  that fix the points 0, 1 and  $\infty$  and the set of smooth complex-valued functions  $\mu$  on  $\mathbb{S}^2$  with  $\|\mu\|_\infty = k < 1$ .

(Here, we identify  $\mathbb{S}^2$  with extended complex plane  $\overline{\mathbb{C}}$ ).

Also, the solution to the Beltrami's eqt depends holomorphically on  $\mu$ . Let  $\{\mu(t)\}$  be a family of Beltrami coefficient depending on a real / complex  $t$ . Let  $\mu(t)(z) = \mu(z) + t\nu(z) + t\varepsilon(t)(z)$  and  $\|\varepsilon(t)\|_\infty \rightarrow 0$  as  $t \rightarrow 0$ . Then: for  $\forall w \in \mathbb{C}$ , for  $z \in \mathbb{C}$   
 $\nu, \varepsilon(t) \in L^\infty(\mathbb{C})$

$f^{M(t)}(w) = f^M(w) + tV(f^M, \nu)(w) + o(|t|)$  locally uniformly on  $\mathbb{C}$

as  $t \rightarrow 0$ , where  $V(f^M, \nu)(w) = \frac{-f^M(w)(f^M(w)-1)}{\pi} \int_{\mathbb{C}} \frac{\nu(z)(f^M(z)-1)^2 dx dy}{f^M(z)(f^M(z)-1)(f^M(z)-f^M(w))}$

Proceed to construct:

$$\begin{array}{ccccccc}
 \tilde{M}_0 & \rightarrow & \tilde{M}_1 & \rightarrow & \tilde{M}_2 & \rightarrow & \dots \rightarrow \tilde{M}_k \rightarrow \dots \rightarrow \tilde{M}_N = \tilde{M} \\
 \updownarrow & & \updownarrow & & \updownarrow & & \updownarrow & & \updownarrow \\
 \tilde{f}^{\tilde{M}_0} & \rightarrow & \tilde{f}^{\tilde{M}_1} & \rightarrow & \tilde{f}^{\tilde{M}_2} & \rightarrow & \dots \rightarrow \tilde{f}^{\tilde{M}_k} \rightarrow \dots \rightarrow \tilde{f}^{\tilde{M}_N} = \tilde{g} \\
 \text{= Id} & & & & & & & & 
 \end{array}$$

Note that the above diagram can be realized iteratively by:

$$\left\{ \begin{array}{l} \tilde{f}^{\tilde{M}_0} = \text{Id} \\ \tilde{f}^{\tilde{M}_{k+1}} = \tilde{f}^{\tilde{M}_k} + \vec{\nabla}(\tilde{f}^{\tilde{M}_k}, \frac{\tilde{M}}{N}), \end{array} \right.$$

where

$$\vec{\nabla}(f^M, v) = - \int_{\mathbb{C}} \frac{f^M(w) (f^M(w) - 1)}{\pi} \left( \frac{v(z) ((f^M)_z(z))^2}{f^M(z) (f^M(z) - 1) (f^M(z) - f^M(w))} \right) dx dy$$

(area)

In discrete case,  $\vec{\nabla}(f^M, v)$  can be discretized as:  $\sum_v k(v, w) \Delta v$

Example: Let  $\phi_1: S_1 \rightarrow D$  and  $\phi_2: S_2 \rightarrow D$  be global conformal parameterizations of  $S_1$  and  $S_2$  respectively.

Want to find  $f: D \rightarrow D$  s.t.  $f$  minimizes:

$$\bar{E}(f) = \int_D (F_1(w) - F_2 \circ f(w))^2 + |M(f)(w)|^2 dw$$

It can be formulated in term of  $\mu$ :

$$\bar{E}(\mu) = \int_D (F_1(w) - F_2 \circ f^\mu(w))^2 + |\mu(w)|^2 dw.$$

To iteratively minimize  $\bar{E}$ , consider:

$$\begin{aligned} \frac{d}{dt} \Big|_{t=0} \bar{E}(\mu + tv) &= \int_D \frac{d}{dt} \Big|_{t=0} (F_1(w) - F_2 \circ f^{\mu+tv}(w))^2 + |\mu(w) + tv(w)|^2 dw \\ &= - \int_D 2 (F_1 - F_2(f^\mu)) \nabla F_2(f^\mu) \frac{d}{dt} \Big|_{t=0} f^{\mu+tv} - 2\mu \cdot v \end{aligned}$$

$$\text{Let } \begin{pmatrix} A \\ B \end{pmatrix} = 2(F_1 - F_2(f^M)) \nabla F_2 ; \quad M = M_1 + iM_2 \equiv \begin{pmatrix} M_1 \\ M_2 \end{pmatrix} ; \quad v = v_1 + i v_2$$

$$\text{Also, we can write: } \overline{\nabla}(f^M, v) \left( = \frac{d}{dt} \Big|_{t=0} f^{M+tv} \right) = \int_D \left( G_1 v_1 + G_2 v_2 \right) dz$$

$$\text{Then: } \frac{d}{dt} \Big|_{t=0} E(M+tv) = - \int_D \left[ \left( \int_D \begin{pmatrix} A \\ B \end{pmatrix} \cdot \begin{pmatrix} G_1 v_1 + G_2 v_2 \\ G_3 v_1 + G_4 v_2 \end{pmatrix} dz \right) - 2 M \cdot v \right] dW$$

$$= - \int_D \left( \int_D \begin{pmatrix} AG_1 + BG_3 \\ AG_2 + BG_4 \end{pmatrix} dW - 2 \begin{pmatrix} M_1 \\ M_2 \end{pmatrix} \right) \cdot \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} dz$$

$\therefore$  The descent direction for  $M_1$  is given by:

$$\frac{dM_1}{dt} = \int_D (AG_1 + BG_3) dW - 2M_1 \quad \text{and}$$

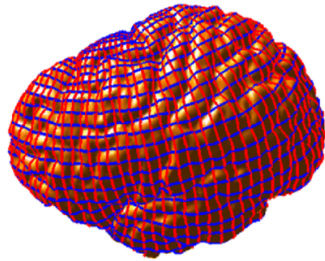
$$\frac{dM_2}{dt} = \int_D (AG_2 + BG_4) dW - 2M_2$$

$\therefore$  We can iteratively optimize the energy  $E$  as follows:

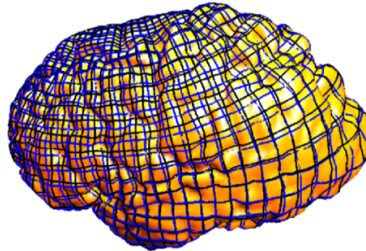
$$\mu^{n+1} = \mu^n + dt \left( \int_D (A^n G_1^n + B^n G_3^n) d\omega - 2\mu_1^n \right) \\ \underbrace{\left( \int_D (A^n G_2^n + B^n G_4^n) d\omega - 2\mu_2^n \right)}_{v^n}$$

$$\begin{array}{ccc} \mu^n & \longrightarrow & \mu^{n+1} \\ \updownarrow & & \updownarrow \\ f^{\mu^n} & & f^{\mu^{n+1}} = f^{\mu^n} + \vec{V}(f^{\mu^n}, v^n) \end{array}$$

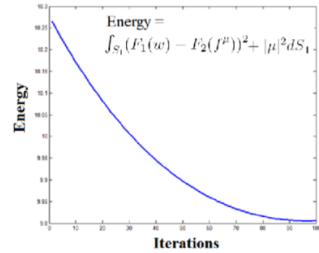
$\therefore$  We get  $(\mu^n, f^{\mu^n})$  iteratively.



(A) Brain 1



(B) Brain 2



(C) Energy in each iteration

Fig. 7

ILLUSTRATION OF BHF OPTIMIZATION SCHEME ON BRAIN SURFACES. THIS EXAMPLE SHOWS THE OPTIMIZATION RESULT OF ATCHING TWO FEATURE FUNCTIONS  $F_1$  AND  $F_2$  ON THE TWO BRAIN SURFACES. THE BLUE GRID REPRESENTS THE INITIAL MAP, WHILE THE BLACK GRID REPRESENTS THE OPTIMIZED MAP.

with  $F_1 \stackrel{\text{def}}{=} 5.2x^2 + 3.3y^2$   
 $F_2 \stackrel{\text{def}}{=} 6.8x^2 + 2.8y$

**Theorem 4.2** (Beltrami holomorphic flow on  $\mathbb{D}$ ) *There is a one-to-one correspondence between the set of quasiconformal diffeomorphisms of  $\mathbb{D}$  that fix the points 0 and 1 and the set of smooth complex-valued functions  $\mu$  on  $\mathbb{D}$  for which  $\|\mu\|_\infty = k < 1$ . Furthermore, the solution  $f^\mu$  depends holomorphically on  $\mu$ . Let  $\{\mu(t)\}$  be a family of Beltrami coefficients depending on a real or complex parameter  $t$ . Suppose also that  $\mu(t)$  can be written in the form*

$$\mu(t)(z) = \mu(z) + tv(z) + t\epsilon(t)(z)$$

for  $z \in \mathbb{D}$ , with suitable  $\mu$  in the unit ball of  $C^\infty(\mathbb{D})$ ,  $v, \epsilon(t) \in L^\infty(\mathbb{D})$  such that  $\|\epsilon(t)\|_\infty \rightarrow 0$  as  $t \rightarrow 0$ . Then for all  $w \in \mathbb{D}$

$$f^{\mu(t)}(w) = f^\mu(w) + tV(f^\mu, v)(w) + o(|t|)$$

locally uniformly on  $\mathbb{D}$  as  $t \rightarrow 0$ , where

$$V(f^\mu, v)(w) = -\frac{f^\mu(w)(f^\mu(w) - 1)}{\pi} \left( \int_{\mathbb{D}} \frac{v(z)((f^\mu)_z(z))^2}{f^\mu(z)(f^\mu(z) - 1)(f^\mu(z) - f^\mu(w))} dx dy \right. \\ \left. + \int_{\mathbb{D}} \frac{\overline{v(z)}(\overline{(f^\mu)_z(z)})^2}{f^\mu(z)(1 - \overline{f^\mu(z)})(1 - \overline{f^\mu(z)}f^\mu(w))} dx dy \right).$$

**Lemma 7.1** Let  $f : \mathbb{D} \rightarrow \mathbb{D}$  be a diffeomorphism of the unit disk fixing 0 and 1 and satisfying the Beltrami equation  $f_{\bar{z}} = \mu f_z$  with  $\mu$  defined on  $\mathbb{D}$ . Let  $\tilde{f}$  be the extension of  $f$  to  $\overline{\mathbb{C}}$  defined as

$$\tilde{f}(z) = \begin{cases} f(z), & \text{if } |z| \leq 1, \\ \frac{1}{f(1/\bar{z})}, & \text{if } |z| > 1. \end{cases}$$

Then  $\tilde{f}$  satisfies the Beltrami equation

$$\tilde{f}_{\bar{z}} = \tilde{\mu} \tilde{f}_z$$

on  $\overline{\mathbb{C}}$ , where the Beltrami coefficient  $\tilde{\mu}$  is defined as

$$\tilde{\mu}(z) = \begin{cases} \mu(z), & \text{if } |z| \leq 1, \\ \frac{z^2}{\bar{z}^2} \overline{\mu(1/\bar{z})}, & \text{if } |z| > 1. \end{cases}$$



## Speeding up Beltrami Holomorphic Flow

Key idea of BHF: Given  $f^\mu$  with Beltrami coefficient  $\mu$ ,  
find  $f^\nu$  with B.C.  $\nu$ .  $\Omega_1 \rightarrow \Omega_2$

Let  $\{\mu_n\}_{n=0}^\infty \ni \mu_0 = \mu$  and  $\mu_\infty = \nu$ . Proceed to modify

$f^\mu$  iteratively:

$$\mu_0 = \mu \rightarrow \mu_1 \rightarrow \dots \rightarrow \mu_n \rightarrow \dots \rightarrow \mu_\infty = \nu$$

$$\begin{array}{c} \uparrow \\ f^{\mu_0} := f^\mu \rightarrow f^{\mu_1} \rightarrow \dots \rightarrow f^{\mu_n} \rightarrow \dots \rightarrow f^{\mu_\infty} = f^\nu \end{array}$$

For a small variation from  $\mu_0$  to  $\mu_1$ , we solve for  $f^{\mu_1} \Rightarrow$

$$f^{\mu_1} = \operatorname{argmin}_{\substack{f \in \text{Diff} \\ \text{Space of} \\ \text{diffeomorphisms}}} \left\{ \left\| \frac{\partial f}{\partial \bar{z}} - \mu_1 \frac{\partial f}{\partial z} \right\|_2^2 \right\}. \text{ Subject} \\ \text{to } f^{\mu_1} \Big|_{\Omega_1} = \Omega_2$$

Let  $f^{\mu_1} = f^{\mu_0} + g_1$ . Reformulate as finding  $g_1 : \Omega_1 \rightarrow \mathbb{R}^2 \ni$

$$(*) \quad g_1 = \operatorname{argmin}_{g: \Omega_1 \rightarrow \mathbb{R}^2} \left\{ \left\| \frac{\partial(f^{\mu_0} + g)}{\partial \bar{z}} - \mu_1 \frac{\partial(f^{\mu_0} + g)}{\partial z} \right\|_2^2 \right\}$$

$$= \operatorname{argmin}_{g: \Omega_1 \rightarrow \mathbb{R}^2} \left\{ \left\| A(\mu_1) g + A(\mu_1) f^{\mu_0} \right\|_2^2 \right\} (*) \text{ subject to boundary constraint.}$$

where  $A(\mu_1) \stackrel{\text{def}}{=} \frac{\partial}{\partial \bar{z}} - \mu_1 \frac{\partial}{\partial z}$ .



Discrete

Least Square Problem

Fast!!

(Instead of computing integration)

**Algorithm 1:** (Beltrami holomorphic flow)

**Input:**  $f^\mu : \Omega_1 \rightarrow \Omega_2$  with BC  $\mu$ , target BC  $\nu$ , threshold  $\epsilon'$

**Output:** Sequence of quasi-conformal maps  $\{f^{\mu_n}\}_{n=1}^\infty$

1. Set  $f^{\mu_0} = f^\mu$ . Solve Equation  $(*)$  to obtain  $g_1$ ;
2. Given  $f^{\mu_n}$ , compute  $\mu_n := \mu(f_n)$  and  $\nu_n := (1 - \epsilon)\mu_n + \epsilon\nu$ ; solve Equation  $(*)$  to obtain  $g_{n+1}$ ; Set  $f_{n+1} := f_n + g_{n+1}$ ;
3. If  $\|\mu_{n+1} - \mu_n\| \geq \epsilon'$ , repeat step 2. Otherwise, stop the iteration.

# ADMM + BHF to solve Diffeomorphism Optimization Problem

Background: Minimize  $\{ E_1(x) + E_2(Ax) \}$  where  $A \in M_{m \times n}(\mathbb{R})$  has full column rank.  $E_1(x)$  usually convex.

Reformulate: (\*) Minimize  $\{ E_1(x) + E_2(y) \}$  Subject to  $Ax = y$ .

Then: the augmented Lagrangian is given by:

$$L(x, y, \lambda, \mu) = E_1(x) + E_2(y) + \lambda^T (Ax - y) + \frac{\mu_k}{2} \|Ax - y\|^2$$

(\*) can be solved by:

$$\begin{cases} (x^{k+1}, y^{k+1}) = \operatorname{argmin} \{ L(x, y, \lambda^k, \mu^k) \} \\ \lambda^{k+1} = \lambda^k + \mu_k (Ax^{k+1} - y^{k+1}) \end{cases} \quad \text{⊗}$$

Different choices of  $\{ \mu_k \}$  has been proposed to ensure convergence.

↑  
Penalty term

⊛ can be difficult to solve.

Alternating direction method with multiplier (ADMM) =

$$\begin{cases} x^{k+1} = \operatorname{argmin}_x \{ L(x, y^k, \lambda^k, \mu_k) \} \\ y^{k+1} = \operatorname{argmin}_y \{ L(x^{k+1}, y, \lambda^k, \mu_k) \} \\ \lambda^{k+1} = \lambda^k + \mu_k (Ax^{k+1} - y^{k+1}) \end{cases}$$

Remark: Once  $\{\lambda^k\}$  and  $\{\mu_k\}$  are carefully chosen, ADMM minimizes in few iterations.

Now, suppose we need to solve Diffeomorphism Optimization Problem:

$$f^* = \operatorname{argmin}_{f \in \text{Diff}} \{ E_1(f) + E_2(\mu(f)) \} \text{ subject to:}$$

$$\|\mu(f^*)\|_\infty \stackrel{\text{def}}{=} \left\| \left( \frac{\partial f^*}{\partial z} \right) / \left( \frac{\partial f^*}{\partial \bar{z}} \right) \right\|_\infty < 1.$$

e.g. Find  $f^*: S_1 \rightarrow S_2$  such that it minimizes:

$$E(f) = \int_{S_1} |\mu(f)|^p + \alpha \int_{S_1} |H_1 - H_2(f)|^2$$

Reformulate: Find  $f^*: S_1 \rightarrow S_2$  and  $v^*: S_1 \rightarrow \mathbb{C} \rightarrow$  mean curvatures

$$(f^*, v^*) = \operatorname{argmin}_{\substack{f \in \text{Diff} \\ \mu \in B}} \{ E_1(f) + E_2(\mu) \} \Rightarrow \begin{cases} \textcircled{1} v^* = \mu(f^*) \\ \textcircled{2} \|v^*\|_\infty < 1. \end{cases}$$

Space of Beltrami coefficient

Augmented Lagrangian:

$$L(f, v, \lambda_{\text{Re}}, \lambda_{\text{Im}}, \rho) = E_1(f) + E_2(v) + \langle \lambda_{\text{Re}}, \text{Re}(v - \mu(f)) \rangle + \langle \lambda_{\text{Im}}, \text{Im}(v - \mu(f)) \rangle + \frac{\rho}{2} \|v - \mu(f)\|_2^2$$

where  $\langle \alpha, \beta \rangle \stackrel{\text{def}}{=} \int_{S_1} \alpha \beta$  and  $\|\alpha\| \stackrel{\text{def}}{=} \left( \int_{S_1} |\alpha|^2 \right)^{1/2}$

Using ADMM:

$$f^{k+1} = \underset{f}{\text{argmin}} \{ L(f, v^k, \lambda_{\text{Re}}^k, \lambda_{\text{Im}}^k, \rho^k) \} \quad - \textcircled{1}$$

$$v^{k+1} = \underset{v}{\text{argmin}} \{ L(f^{k+1}, v, \lambda_{\text{Re}}^k, \lambda_{\text{Im}}^k, \rho^k) \} \quad - \textcircled{2}$$

$\lambda_{\text{Re}}^k$ ,  $\lambda_{\text{Im}}^k$  and  $\rho^k$  are updated as follows:

if  $\|v^{k+1} - \mu(f^{k+1})\|_2 < \eta_k$ , update:

$$\lambda_{\text{Re}}^{k+1} = \lambda_{\text{Re}}^k + \rho_k \text{Re}(v^{k+1} - \mu(f^{k+1}))$$

$$\lambda_{\text{Im}}^{k+1} = \lambda_{\text{Im}}^k + \rho_k \text{Im}(v^{k+1} - \mu(f^{k+1}))$$

$$\rho_{k+1} = \rho_k$$

if  $\|v^{k+1} - u(f^{k+1})\| \geq \eta_k$ , update:

$$\lambda_{\text{Re}}^{k+1} = \lambda_{\text{Re}}^k ; \lambda_{\text{Im}}^{k+1} = \lambda_{\text{Im}}^k$$

$$\rho_{k+1} = \rho_k (1 + \gamma_k)$$

( $\eta_k$  is chosen to be  $\downarrow$ )  
( $\gamma_k$  can be constant)

Solving subproblem ② involving  $v$ :

Very often, Euler-Lagrange eqt of  $E_2(u)$  is an elliptic PDE. e.g.  $\int_{\Omega} |\nabla u|^2 + |u|^2$ . Then, E-L eqt of

② can be written as:

$$\Delta u - 2u - \lambda_{\text{Re}}^k \text{Re}(v - u(f^{k+1})) - \lambda_{\text{Im}}^k \text{Im}(v - u(f^{k+1})) - \rho_k (v - u(f^{k+1})) = 0$$

Discretize  $\rightsquigarrow$  Sparse Symmetric positive definite linear system.

Solving subproblem ① involving  $f: \tilde{v}$

$$f^{k+1} = \underset{f}{\operatorname{argmin}} \left\{ E_1(f) + \langle \lambda_{\operatorname{Re}}, \operatorname{Re}(v - u(f)) \rangle + \langle \lambda_{\operatorname{Im}}, \operatorname{Im}(v - u(f)) \rangle + \frac{\rho}{2} \|v - u(f)\|_2^2 \right\}$$

$E_1(f)$  can be minimized using gradient descent algorithm.

e.g.  $E_1(f) = \|I_1 - I_2(f)\|_2^2$

Curvatures on  $S_1$  and  $S_2$

Then: descent direction  $\vec{v}_1 = 2(I_1 - I_2(f)) \nabla f$

Last three terms can be minimized over B.C. to get a descent direction  $\partial \tilde{v} =$

$$\partial \tilde{v} = -\lambda_{\operatorname{Re}} - \lambda_{\operatorname{Im}} + \rho(v - u(f))$$

After few iteration, we get a new B.C.  $\tilde{v}$ .



We can find the associated QC map  $\tilde{f} = f^k + \vec{V}_2 \Rightarrow$

$$\frac{\partial \tilde{f}}{\partial \bar{z}} = \vec{V} \frac{\partial f}{\partial z}$$

Overall, the descent direction to optimize :

$$\frac{df^k}{dt} = \vec{V}_1 + \vec{V}_2$$