

Assignment Five: Geometric Optimal Transport Map

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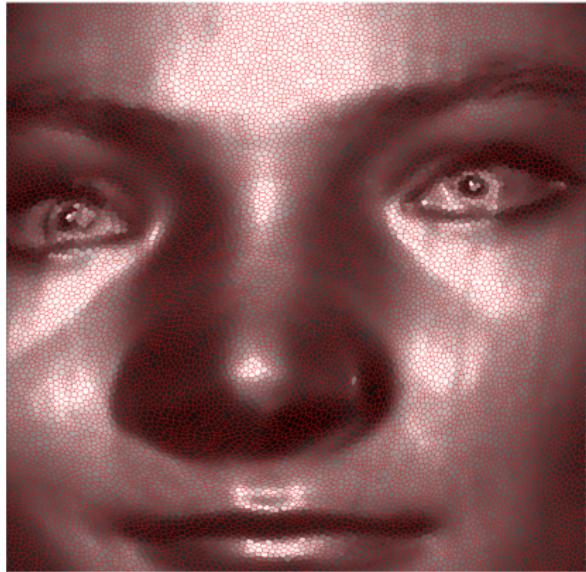
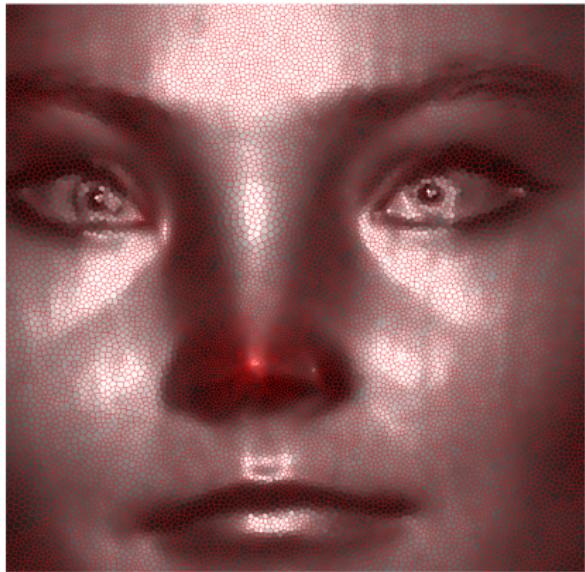
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Computational Results



Computational Results



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How to eliminate mode collapse?

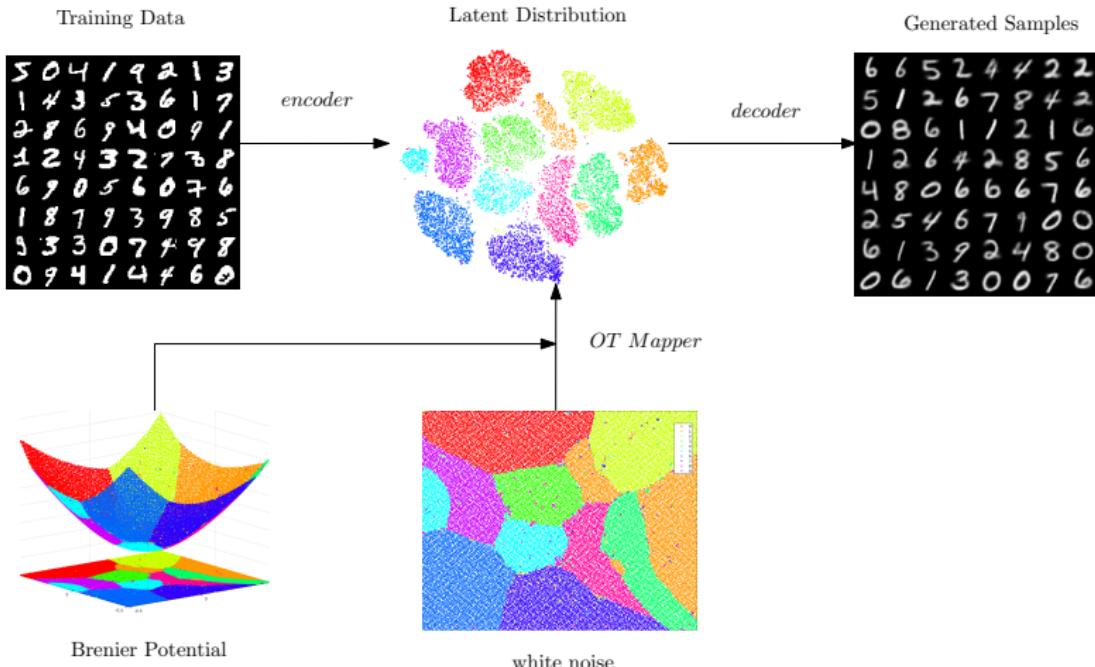


Figure: Geometric Generative Model.

Convex Geometric View

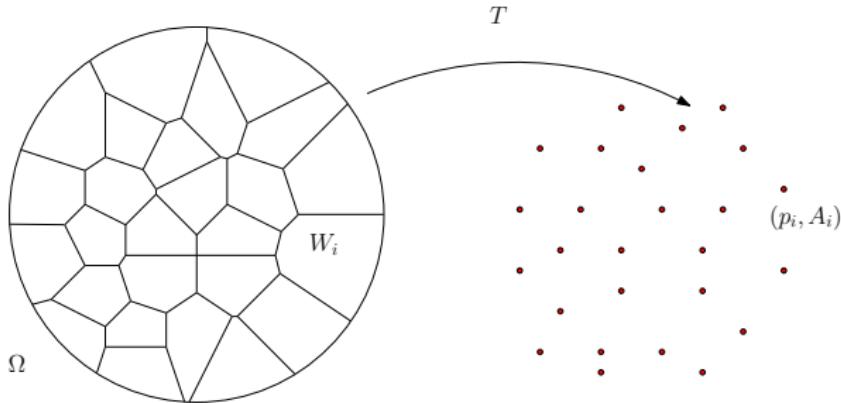
Monge-Ampère Equation

Problem (Brenier)

Given (Ω, μ) and (Σ, ν) and the cost function $c(x, y) = \frac{1}{2}|x - y|^2$, the optimal transportation map $T : \Omega \rightarrow \Sigma$ is the gradient map of the Brenier potential $u : \Omega \rightarrow \mathbb{R}$, which satisfies the Monge-Ampère equation,

$$\det \left(\frac{\partial^2 u(x)}{\partial x_i \partial x_j} \right) = \frac{f(x)}{g \circ \nabla u(x)}$$

Semi-Discrete Optimal Transportation Problem

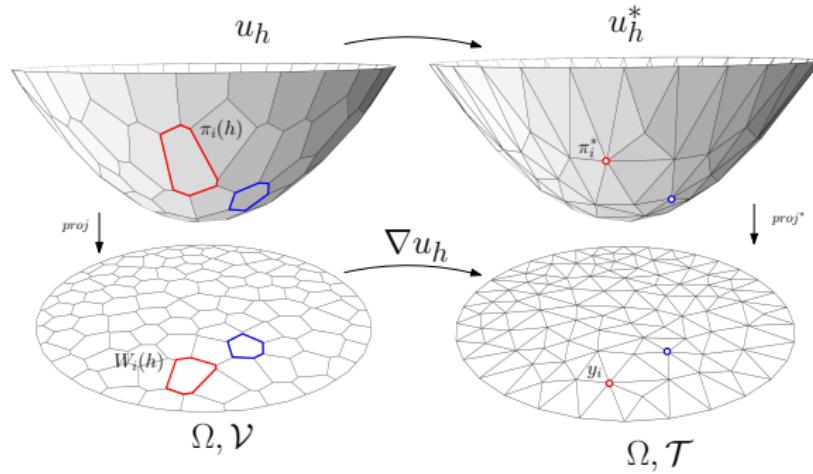


Problem (Semi-discrete OT)

Given a compact convex domain Ω in \mathbb{R}^d , and p_1, p_2, \dots, p_k and weights $w_1, w_2, \dots, w_k > 0$, find a transport map $T : \Omega \rightarrow \{p_1, \dots, p_k\}$, such that $\text{vol}(T^{-1}(p_i)) = w_i$, so that T minimizes the transportation cost:

$$\mathcal{C}(T) := \frac{1}{2} \int_{\Omega} |x - T(x)|^2 dx$$

Semi-Discrete Optimal Transportation Problem



According to Brenier theorem, there will be a piecewise linear convex function $u : \Omega \rightarrow \mathbb{R}$, the gradient map gives the optimal transport map.

Alexandrov Theorem

Theorem (Alexandrov 1950)

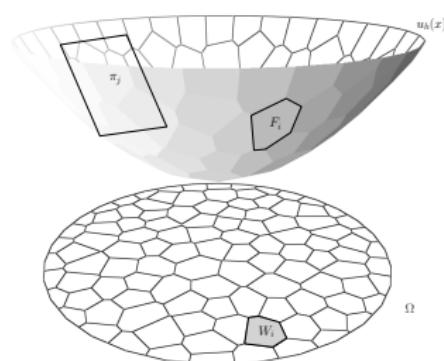
Given Ω compact convex domain in \mathbb{R}^n ,
 p_1, \dots, p_k distinct in \mathbb{R}^n , $A_1, \dots, A_k > 0$,
such that $\sum A_i = \text{Vol}(\Omega)$, there exists PL
convex function

$$f(\mathbf{x}) := \max\{\langle \mathbf{x}, \mathbf{p}_i \rangle + h_i | i = 1, \dots, k\}$$

unique up to translation such that

$$\text{Vol}(W_i) = \text{Vol}(\{\mathbf{x} | \nabla f(\mathbf{x}) = \mathbf{p}_i\}) = A_i.$$

Alexandrov's proof is topological, not variational. It has been open for years to find a constructive proof.



Variational Proof

Theorem (Gu-Luo-Sun-Yau 2013)

Ω is a compact convex domain in \mathbb{R}^n , y_1, \dots, y_k distinct in \mathbb{R}^n , μ a positive continuous measure on Ω . For any $\nu_1, \dots, \nu_k > 0$ with $\sum \nu_i = \mu(\Omega)$, there exists a vector (h_1, \dots, h_k) so that

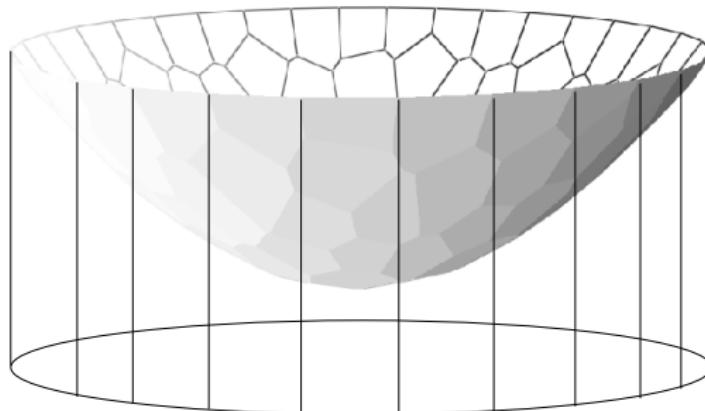
$$u(\mathbf{x}) = \max\{\langle \mathbf{x}, \mathbf{p}_i \rangle + h_i\}$$

satisfies $\mu(W_i \cap \Omega) = \nu_i$, where $W_i = \{\mathbf{x} | \nabla f(\mathbf{x}) = \mathbf{p}_i\}$. Furthermore, \mathbf{h} is the maximum point of the concave function

$$E(\mathbf{h}) = \sum_{i=1}^k \nu_i h_i - \int_0^{\mathbf{h}} \sum_{i=1}^k w_i(\eta) d\eta_i,$$

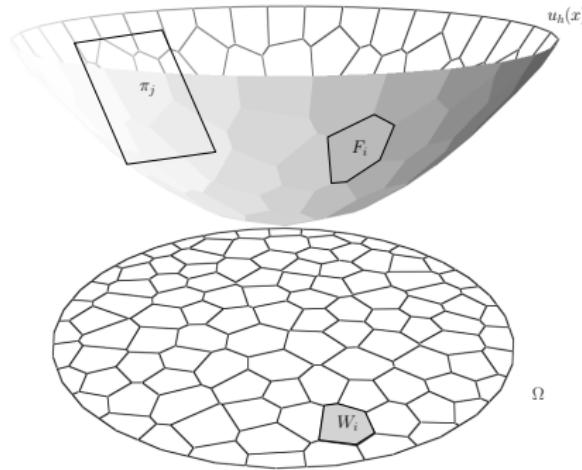
where $w_i(\eta) = \mu(W_i(\eta) \cap \Omega)$ is the μ -volume of the cell.

Geometric Interpretation



One can define a cylinder through $\partial\Omega$, the cylinder is truncated by the xy-plane and the convex polyhedron. The energy term $\int^h \sum w_i(\eta) d\eta_i$ equals to the volume of the truncated cylinder.

Computational Algorithm

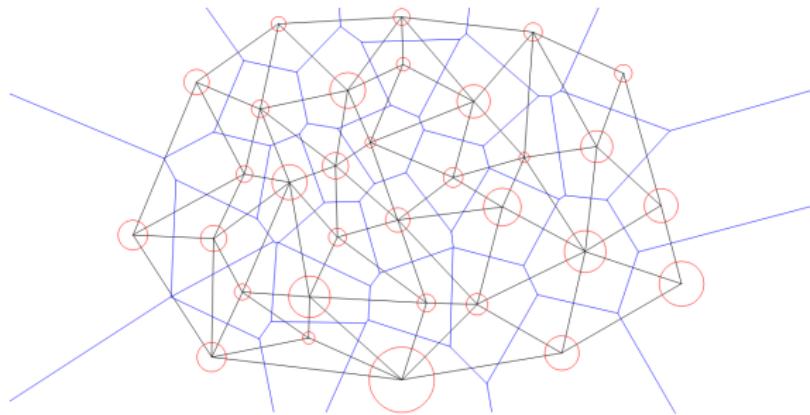


Definition (Alexandrov Potential)

The concave energy is

$$E(h_1, h_2, \dots, h_k) = \sum_{i=1}^k \nu_i h_i - \int_{\mathbf{0}}^{\mathbf{h}} \sum_{j=1}^k w_j(\eta) d\eta_j,$$

Computational Algorithm



The Hessian of the energy is the length ratios of edge and dual edges,

$$\frac{\partial w_i}{\partial h_j} = - \frac{|\mathbf{e}_{ij}|}{|\bar{\mathbf{e}}_{ij}|}$$

Optimal Transport Map

Input: A set of distinct points $P = \{p_1, p_2, \dots, p_k\}$, and the weights $\{A_1, A_2, \dots, A_k\}$; A convex domain Ω , $\sum A_j = \text{Vol}(\Omega)$;

Output: The optimal transport map $T : \Omega \rightarrow P$

- ① Scale and translate P , such that $P \subset \Omega$;
- ② Initialize $\mathbf{h}^0 \leftarrow \frac{1}{2}(|p_1|^2, |p_2|^2, \dots, |p_k|^2)^T$;
- ③ Compute the Brenier potential $u(\mathbf{h}^k)$ (envelope of π_i 's) and its Legendre dual $u^*(\mathbf{h}^k)$ (convex hull of π_i^* 's);
- ④ Project the Brenier potential and Legendre dual to obtain weighted Delaunay triangulation $\mathcal{T}(\mathbf{h}^k)$ and power diagram $\mathcal{D}(\mathbf{h}^k)$;

Optimal Transport Map

- ⑤ Compute the gradient of the energy

$$\nabla E(\mathbf{h}) = (A_1 - w_1(\mathbf{h}), A_2 - w_2(\mathbf{h}), \dots, A_k - w_k(\mathbf{h}))^T.$$

- ⑥ If $\|E(\mathbf{h}^k)\|$ is less than ε , then return $T = \nabla u(\mathbf{h}^k)$;

- ⑦ Compute the Hessian matrix of the energy

$$\frac{\partial w_i(\mathbf{h})}{\partial h_j} = -\frac{|e_{ij}|}{|\bar{e}_{ij}|}, \quad \frac{\partial w_i}{\partial h_i} = -\sum_j \frac{\partial w_i(\mathbf{h})}{\partial h_j}.$$

- ⑧ Solve linear system

$$\nabla E(\mathbf{h}) = \text{Hess}(\mathbf{h}^k) \mathbf{d};$$

Optimal Transport Map

- ⑨ Set the step length $\lambda \leftarrow 1$;
- ⑩ Construct the convex hull $\text{Conv}(\mathbf{h}^k + \lambda \mathbf{d})$;
- ⑪ if there is any empty power cell, $\lambda \leftarrow \frac{1}{2}\lambda$, repeat step 3 and 4, until all power cells are non-empty;
- ⑫ set $\mathbf{h}^{k+1} \leftarrow \mathbf{h}^k + \lambda \mathbf{d}$;
- ⑬ Repeat step 3 through 14.

Optimal Transportation Map



Figure: Optimal transportation map.

Optimal Transportation Map

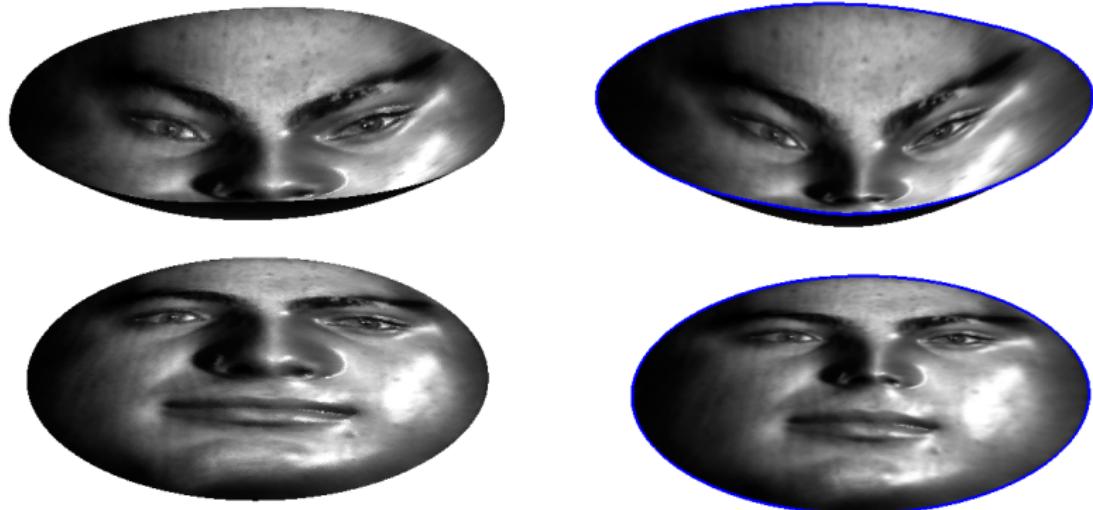


Figure: Optimal transportation map.

Instruction

Dependencies

- ① 'detri2', a mesh generation library, written by Dr. Hang Si.
- ② 'MeshLib', a mesh library based on halfedge data structure.
- ③ 'freeglut', a free-software/open-source alternative to the OpenGL Utility Toolkit (GLUT) library.

Directory Structure

- ot_2d/include, the header files for optimal transport;
- ot_2d/src, the source files for optimal transport.
- data, Some models.
- CMakeLists.txt, CMake configuration file.
- resources, Some resources needed.
- 3rdparty, MeshLib and freeglut libraries.

Configuration

Before you start, read README.md carefully, then go through the following procedures, step by step.

- ① Install [CMake](<https://cmake.org/download/>).
- ② Download the source code of the C++ framework.
- ③ Configure and generate the project for Visual Studio.
- ④ Open the .sln using Visual Studio, and compile the solution.
- ⑤ Finish your code in your IDE.
- ⑥ Run the executable program.

3. Configure and generate the project

- ① open a command window
- ② cd Assignment_6_skeleton
- ③ mkdir build
- ④ cd build
- ⑤ cmake ..
- ⑥ open CCGHomework.sln inside the build directory.

5. Finish your code in your IDE

- You need to modify the file: OT.cpp and CDomainOptimalTransport.cpp
- search for comments “insert your code here”
- Modify functions:
 - ① CDomainTransport::_newton(COMTMesh * pInput, COMTMesh * pOutput)
 - ② CBaseOT::_update_direction(COMTMesh* pMesh)
 - ③ CBaseOT::_compute_hessian_matrix(COMTMesh& mesh, Eigen::SparseMatrix& hessian)

6. Run the executable program

Dynamic Linking Libraries

Copy detri2.dll and detri2d.dll from 3rdparty/detri2/lib/windows to build/ot_2d/; Libraries and dlls for Linux and MAC are also available.

Command

Command line:

OT2d.exe girl.m

All the data files are in the data folder, all the texture images are in the textures folder.