

## Warped-Seam Geometry and Canalization as Gromov–Hausdorff Collapse

LARS RÖNNBÄCK\*

*Department of Computer and Systems Sciences, Stockholm University, Stockholm, Sweden*

\*Corresponding author: [lars@uptochange.com](mailto:lars@uptochange.com)

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Waddington’s epigenetic landscape is often drawn as a “ball rolling” picture; here we give a concrete metric realization directly on gene-expression space  $\mathbb{R}^n$ . Given a smooth nonnegative scalar field  $s$  (“landscape height”) and an admissible warping function  $\phi$  with  $\phi(0) = 0$ , we deform the Euclidean metric conformally by the factor  $\phi(s)^2$ . The resulting length metric collapses distances inside each basin of attraction. Under Morse, separation, and properness assumptions on  $s$ , we prove that the compact sublevel sets  $X_\varepsilon = \{s \leq \varepsilon\}$  converge in the Gromov–Hausdorff sense to a finite discrete metric space with one point per global minimum, and establish a sharp two-sided bound  $\text{diam}(X_\varepsilon^{(i)}, d_\phi) = \Theta(\varepsilon^{\alpha+1/2})$  for power-law warping  $\phi(s) = s^\alpha$ . We extend the framework to Morse–Bott landscapes, where the zero set of  $s$  is a union of compact connected submanifolds—modelling attracting cell-type manifolds rather than isolated fixed points—and show that each connected component still collapses to a single point under  $d_\phi$ . We connect the warped metric to the Freidlin–Wentzell quasi-potential for the gradient stochastic differential equation  $dX = -\nabla s(X) dt + \sqrt{2T} dW$ , showing that the quasi-potential is itself a warped length metric with position-dependent weight  $|\nabla s|$ , and that the dynamically natural choice  $\phi(s) = \sqrt{s}$  arises from the local quadratic structure of  $s$  near its minima. Finally, we illustrate the theory on a two-dimensional toggle-switch landscape modelling mutual gene repression, computing collapse rates and inter-basin distances explicitly.

**Keywords:** canalization; Waddington landscape; Gromov–Hausdorff convergence; Morse theory; Freidlin–Wentzell theory.

### 1. Introduction and Overview

In developmental biology, *canalization* refers informally to robustness: many nearby states lead to the same outcome. Waddington’s epigenetic landscape (Waddington, 1957) depicts this as valleys separated by ridges, and a large body of quantitative work—from quasi-potential reconstructions in gene-regulatory networks (Wang et al., 2011; Zhou et al., 2012) to single-cell trajectory inference (Rizvi et al., 2017; Schiebinger et al., 2019)—has sought to make this picture precise.

This paper isolates a minimal geometric statement capturing one aspect of that picture: as we restrict to low landscape height  $s \leq \varepsilon$ , each basin becomes metrically small, while distinct basins remain at a finite distance from each other. The mechanism is simple: define a weighted length functional where the cost of motion is multiplied by a factor  $\phi(s)$  that vanishes at the global minima of  $s$ . Inside a low sublevel set, motion can “hug” the minimum and become cheap; any path connecting distinct minima must cross a ridge where  $s$  is bounded away from 0, forcing positive length.

We call the resulting structure a *warped-seam geometry*: the *seam* is the scalar field  $s$  itself—so named because it encodes how the space is stitched together, prescribing where basins meet and where ridges separate them. The *warping* is the rule  $\phi$  that converts the seam into a genuine metric via  $g_\phi = \phi(s)^2 g_F$ . Together, the seam tells us *what* the geometric structure is; the warping tells us *how strongly* it is felt (see Figure 1).

Gromov–Hausdorff convergence of degenerating metric spaces is a well-developed subject. The Cheeger–Gromov–Fukaya theory (Cheeger and Gromov, 1986; Fukaya, 1987) studies collapsing of Riemannian manifolds under curvature bounds—a substantially deeper phenomenon than what occurs here. Our collapsing mechanism is more elementary: the conformal factor  $\phi(s)$  simply vanishes at the minima, making basins metrically small without any curvature control. Our contribution is not the collapsing mechanism itself but its application to the canalization setting: we identify the biologically meaningful assumptions under which GH collapse occurs, establish sharp rates, connect the warped metric to stochastic dynamics, and provide explicit computations for gene-regulatory models.

**Contributions.** Beyond the basic Gromov–Hausdorff collapse theorem (Theorem 6.2), we make four additional contributions:

- (1) *Sharp collapse rate* (Proposition 7.2). We prove a matching lower bound on the diameter, yielding  $\text{diam}(X_\varepsilon^{(i)}, d_\phi) = \Theta(\varepsilon^{\alpha+1/2})$  for  $\phi(s) = s^\alpha$ .
- (2) *Morse–Bott generalization* (Theorem 8.4). When the zero set  $\{s = 0\}$  is a disjoint union of compact submanifolds—the biologically relevant case where a cell type is a manifold of expression states—the same GH collapse holds.
- (3) *Freidlin–Wentzell connection* (Section 9). We show the quasi-potential of the gradient SDE  $dX = -\nabla s dt + \sqrt{2T} dW$  is itself a warped length metric, and that  $\phi(s) = \sqrt{s}$  is the dynamically natural warping.
- (4) *Toggle-switch example* (Section 10.2). We work out a concrete two-dimensional bistable landscape modelling mutual gene repression, computing inter-basin distances and collapse rates explicitly.

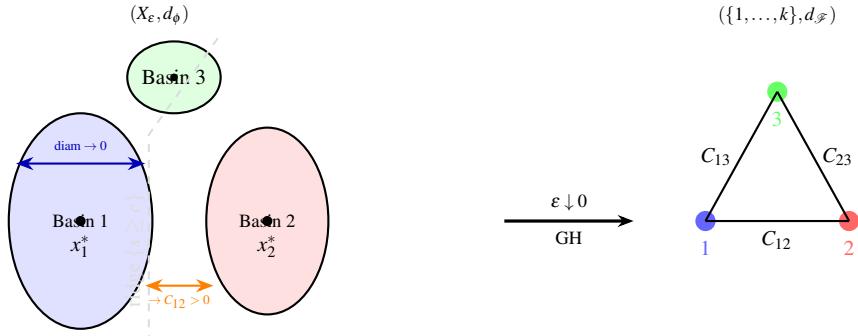


Figure 1. Schematic of Gromov–Hausdorff collapse. Left: the sublevel set  $X_\varepsilon$  with three basins separated by ridges (dashed). Under the warped metric  $d_\phi$ , each basin has diameter  $\rightarrow 0$  while inter-basin distances converge to finite constants  $C_{ij}$ . Right: the limit discrete metric space.

## 2. Geometric Framework

Let  $F = \mathbb{R}^n$  with Euclidean metric  $g_F$ . Let  $s : F \rightarrow [0, \infty)$  be smooth.

**Definition 2.1** (Admissible warping function) *A function  $\phi : [0, \infty) \rightarrow [0, \infty)$  is admissible if:*

- (i)  $\phi$  is continuous on  $[0, \infty)$  and smooth on  $(0, \infty)$ ,
- (ii)  $\phi$  is strictly increasing,
- (iii)  $\phi(0) = 0$ , and
- (iv)  $\phi(s) > 0$  for  $s > 0$ .

**Definition 2.2** (Conformal length metric) *Define the conformal (possibly degenerate) tensor*

$$g_\phi := \phi(s(x))^2 g_F.$$

*The induced length metric is*

$$d_\phi(x, y) = \inf_{\gamma} \int_0^1 \phi(s(\gamma(u))) |\gamma'(u)| du,$$

*where the infimum ranges over piecewise  $C^1$  curves  $\gamma : [0, 1] \rightarrow \mathbb{R}^n$  with  $\gamma(0) = x$  and  $\gamma(1) = y$ .*

**Remark 2.3** (Degeneracy at minima) *Since  $\phi(0) = 0$ , the tensor  $g_\phi$  degenerates on the zero set  $\{s = 0\}$ . Nevertheless the definition above yields an extended length metric on  $\mathbb{R}^n$ . Under the separation assumption below, distinct minima have strictly positive  $d_\phi$ -distance.*

For subsets  $A, B \subset \mathbb{R}^n$  we write

$$d_\phi(A, B) := \inf\{d_\phi(a, b) : a \in A, b \in B\}.$$

### 3. Structural Assumptions

We assume:

- (A1)  $s$  is Morse.
- (A2)  $s$  has exactly  $k$  nondegenerate global minima  $x_1^*, \dots, x_k^*$  with  $s(x_i^*) = 0$ .
- (A3) (Ridge separation) There exists  $c > 0$  such that the connected components of  $\{s < c\}$  containing distinct minima have disjoint closures.
- (A4) (Properness)  $s(x) \rightarrow \infty$  as  $|x| \rightarrow \infty$ .

Properness implies all sublevel sets are compact. For  $\varepsilon > 0$  define

$$X_\varepsilon := \{x \in \mathbb{R}^n : s(x) \leq \varepsilon\},$$

and for  $0 < \varepsilon < c$  let  $X_\varepsilon^{(i)}$  denote the component containing  $x_i^*$ .

### 4. Local quadratic control and shrinkage

**Lemma 4.1** (Two-sided quadratic control near minima) *For each  $i$  there exist constants  $r_i > 0$  and  $0 < \lambda_i \leq \Lambda_i$  such that for all  $|x - x_i^*| < r_i$ ,*

$$\frac{\lambda_i}{2} |x - x_i^*|^2 \leq s(x) \leq \frac{\Lambda_i}{2} |x - x_i^*|^2.$$

*Proof* Since  $x_i^*$  is a nondegenerate minimum,  $\text{Hess } s(x_i^*)$  is positive definite. By continuity of  $\text{Hess } s$  there is a neighborhood where the eigenvalues of  $\text{Hess } s(x)$  lie in  $[\lambda_i, \Lambda_i]$ . Integrating the second-order Taylor formula along the segment from  $x_i^*$  to  $x$  gives the stated two-sided bounds.  $\square$

**Lemma 4.2** (Sublevel shrinkage in Euclidean diameter) *For each  $i$  there exists  $C_i > 0$  such that for all sufficiently small  $\varepsilon$ ,*

$$\text{diam}(X_\varepsilon^{(i)}, g_F) \leq C_i \sqrt{\varepsilon}.$$

*Proof* By Lemma 4.1,  $s(x) \leq \varepsilon$  implies  $|x - x_i^*| \leq \sqrt{2\varepsilon/\lambda_i}$ . Hence  $X_\varepsilon^{(i)}$  is contained in a Euclidean ball of that radius, and its Euclidean diameter is  $O(\sqrt{\varepsilon})$ .  $\square$

**Lemma 4.3** (Distance to the minimum) *For each  $i$  there exist  $\varepsilon_i > 0$  and  $K_i > 0$  such that for all  $0 < \varepsilon \leq \varepsilon_i$  and all  $x \in X_\varepsilon^{(i)}$ ,*

$$d_\phi(x, x_i^*) \leq K_i \phi(\kappa_i \varepsilon) \sqrt{\varepsilon},$$

where  $\kappa_i := \Lambda_i/\lambda_i$  comes from Lemma 4.1. Consequently,  $\sup_{x \in X_\varepsilon^{(i)}} d_\phi(x, x_i^*) \rightarrow 0$  as  $\varepsilon \downarrow 0$ .

*Proof* Take  $\varepsilon$  small enough that  $X_\varepsilon^{(i)} \subset B(x_i^*, r_i)$  with  $r_i$  from Lemma 4.1. Fix  $x \in X_\varepsilon^{(i)}$  and consider the Euclidean segment  $\gamma(t) = x_i^* + t(x - x_i^*)$ .

For  $t \in [0, 1]$ , Lemma 4.1 gives

$$s(\gamma(t)) \leq \frac{\Lambda_i}{2} t^2 |x - x_i^*|^2 \leq \frac{\Lambda_i}{\lambda_i} \varepsilon,$$

since  $|x - x_i^*|^2 \leq 2\varepsilon/\lambda_i$ . Thus  $\phi(s(\gamma(t))) \leq \phi((\Lambda_i/\lambda_i)\varepsilon) = \phi(\kappa_i\varepsilon)$  and

$$d_\phi(x, x_i^*) \leq \int_0^1 \phi(s(\gamma(t))) |\gamma'(t)| dt \leq \phi(\kappa_i\varepsilon) |x - x_i^*|.$$

Using  $|x - x_i^*| \leq \sqrt{2\varepsilon/\lambda_i}$  yields the bound with a constant  $K_i$ .  $\square$

**Remark 4.4** (When  $\phi(\kappa\varepsilon) \asymp \phi(\varepsilon)$ ) *For common choices such as  $\phi(s) = s^\alpha$  (or more generally  $\phi$  regularly varying at 0), one has  $\phi(\kappa\varepsilon) \asymp \phi(\varepsilon)$  for each fixed  $\kappa > 0$ . In that case Lemma 4.3 can be stated with  $\phi(\varepsilon)$  up to a change in constants.*

## 5. Inter-basin distances

Fix  $c > 0$  from (A3). For each minimum define the basin at height  $c$ :

$$A_i(c) := \text{the connected component of } \{s < c\} \text{ containing } x_i^*.$$

Define the Euclidean gap

$$w_{ij} := \text{dist}_{g_F}(\overline{A_i(c)}, \overline{A_j(c)}).$$

By (A3) and compactness of the closures,  $w_{ij} > 0$  for  $i \neq j$ .

**Definition 5.1** (Inter-basin constants) *For  $i \neq j$  define*

$$C_{ij} := d_\phi(x_i^*, x_j^*).$$

**Lemma 5.2** (Positivity and finiteness) *For  $i \neq j$  we have  $0 < C_{ij} < \infty$ .*

*Proof* Finiteness follows because  $\mathbb{R}^n$  is path connected and  $\phi \circ s$  is finite everywhere.

For positivity, any curve from  $x_i^*$  to  $x_j^*$  must leave  $A_i(c)$  and enter  $A_j(c)$ , hence must intersect the closed set  $\{s \geq c\}$ . Along the portion of the curve in  $\{s \geq c\}$  we have  $\phi(s) \geq \phi(c)$ . Moreover the curve must cross the Euclidean gap between the disjoint closed sets  $\overline{A_i(c)}$  and  $\overline{A_j(c)}$ , which forces Euclidean arclength at least  $w_{ij}$ . Therefore  $C_{ij} \geq \phi(c)w_{ij} > 0$ .  $\square$

**Lemma 5.3** (Distance convergence) *For  $i \neq j$ ,*

$$d_\phi(X_\varepsilon^{(i)}, X_\varepsilon^{(j)}) \rightarrow C_{ij} \quad (\varepsilon \downarrow 0).$$

*Proof* For any  $x \in X_\varepsilon^{(i)}$  and  $y \in X_\varepsilon^{(j)}$ , the triangle inequality gives

$$|d_\phi(x, y) - d_\phi(x_i^*, x_j^*)| \leq d_\phi(x, x_i^*) + d_\phi(y, x_j^*).$$

Taking infimum over  $x, y$  yields

$$|d_\phi(X_\varepsilon^{(i)}, X_\varepsilon^{(j)}) - C_{ij}| \leq \sup_{x \in X_\varepsilon^{(i)}} d_\phi(x, x_i^*) + \sup_{y \in X_\varepsilon^{(j)}} d_\phi(y, x_j^*),$$

which tends to 0 by Lemma 4.3.  $\square$

## 6. Canalization as Gromov–Hausdorff collapse

**Lemma 6.1** (Intra-basin collapse) *For each  $i$  there exists  $D_i > 0$  such that for all sufficiently small  $\varepsilon$ ,*

$$\text{diam}(X_\varepsilon^{(i)}, d_\phi) \leq D_i \phi(\kappa_i \varepsilon) \sqrt{\varepsilon}.$$

*Proof* For  $x, x' \in X_\varepsilon^{(i)}$  we estimate through the minimum:  $d_\phi(x, x') \leq d_\phi(x, x_i^*) + d_\phi(x_i^*, x')$ . Taking the supremum over  $x, x'$  and applying Lemma 4.3 gives the claim.  $\square$

**Theorem 6.2** (Canalization as GH collapse) *Let  $d_{\mathcal{F}}$  be the metric on  $\{1, \dots, k\}$  given by  $d_{\mathcal{F}}(i, i) = 0$  and  $d_{\mathcal{F}}(i, j) = C_{ij}$  for  $i \neq j$ . Then*

$$(X_\varepsilon, d_\phi) \xrightarrow{\text{GH}} (\{1, \dots, k\}, d_{\mathcal{F}}) \quad (\varepsilon \downarrow 0).$$

*Proof* Fix  $\delta > 0$ . Choose  $\varepsilon$  small such that for all  $i$ ,

$$\text{diam}(X_\varepsilon^{(i)}, d_\phi) < \delta/6$$

by Lemma 6.1, and for all  $i \neq j$ ,

$$|d_\phi(X_\varepsilon^{(i)}, X_\varepsilon^{(j)}) - C_{ij}| < \delta/6$$

by Lemma 5.3.

Define a correspondence  $\mathcal{R} \subset X_\varepsilon \times \{1, \dots, k\}$  by  $(x, i) \in \mathcal{R}$  iff  $x \in X_\varepsilon^{(i)}$ .

If  $(x, i), (y, j) \in \mathcal{R}$  and  $i = j$  then  $d_{\mathcal{F}}(i, j) = 0$  while  $d_\phi(x, y) \leq \text{diam}(X_\varepsilon^{(i)}, d_\phi) < \delta/6$ . If  $i \neq j$ , then

$$|d_\phi(x, y) - d_\phi(X_\varepsilon^{(i)}, X_\varepsilon^{(j)})| \leq \text{diam}(X_\varepsilon^{(i)}, d_\phi) + \text{diam}(X_\varepsilon^{(j)}, d_\phi) < \delta/3,$$

and so

$$|d_\phi(x, y) - d_{\mathcal{F}}(i, j)| < \delta/3 + \delta/6 = \delta/2.$$

Thus the distortion of  $\mathcal{R}$  is  $< \delta$ , which implies  $d_{\text{GH}}((X_\varepsilon, d_\phi), (\{1, \dots, k\}, d_{\mathcal{F}})) < \delta/2$ .  $\square$

## 7. Collapse rate

Near  $x_i^*$ ,

$$s(x) = \frac{1}{2}(x - x_i^*)^\top H_i(x - x_i^*) + O(|x - x_i^*|^3),$$

where  $H_i = \text{Hess } s(x_i^*)$  is positive definite. Let  $\lambda_{\min}$  be the smallest eigenvalue of  $H_i$ .

Lemma 6.1 yields the scaling

$$\text{diam}(X_\varepsilon^{(i)}, d_\phi) \leq C \phi(\kappa_i \varepsilon) \sqrt{\varepsilon / \lambda_{\min}} \quad (\varepsilon \downarrow 0)$$

for a constant  $C$  depending on the local quadratic bounds.

If  $\phi(s) = s^\alpha$  with  $\alpha > 0$ , then

$$\text{diam}(X_\varepsilon^{(i)}, d_\phi) = O(\varepsilon^{\alpha+1/2}).$$

**Remark 7.1** (What controls the rate) *The exponent  $\alpha + \frac{1}{2}$  combines how quickly motion becomes inexpensive near  $s = 0$  (encoded by  $\phi$ ) and how quickly the basin radius shrinks (encoded by the quadratic model of  $s$ ). Constants depend on the local condition number  $\Lambda_i/\lambda_i$  from Lemma 4.1.*

The upper bound is in fact sharp. We now establish a matching lower bound.

**Proposition 7.2** (Matching lower bound on diameter) *Under assumptions (A1)–(A4), for each  $i$  there exists  $c_i > 0$  such that for all sufficiently small  $\varepsilon$ ,*

$$\text{diam}(X_\varepsilon^{(i)}, d_\phi) \geq 2 \int_0^{\sqrt{2\varepsilon/\Lambda_i}} \phi\left(\frac{\lambda_i}{2} u^2\right) du.$$

In particular, for  $\phi(s) = s^\alpha$  with  $\alpha > 0$ ,

$$\text{diam}(X_\varepsilon^{(i)}, d_\phi) = \Theta(\varepsilon^{\alpha+1/2}) \quad (\varepsilon \downarrow 0).$$

*Proof* Let  $e$  be any unit eigenvector of  $H_i = \text{Hess } s(x_i^*)$  and set  $R = \sqrt{2\varepsilon/\Lambda_i}$ , where  $\Lambda_i$  is the largest eigenvalue of  $H_i$ . Define  $x_\pm = x_i^* \pm Re$ . By the upper quadratic bound (Lemma 4.1),

$$s(x_\pm) \leq \frac{\Lambda_i}{2} R^2 = \varepsilon,$$

so  $x_\pm \in X_\varepsilon^{(i)}$  for  $\varepsilon$  small enough that  $R < r_i$ .

*Projection argument.* Write  $\pi(y) = \langle y - x_i^*, e \rangle$  for the signed distance from  $x_i^*$  along  $e$ . For any piecewise  $C^1$  curve  $\gamma$  from  $x_+$  to  $x_-$  we have  $|\gamma'| \geq |\pi'(\gamma)|$ . Moreover, the lower quadratic bound gives  $s(\gamma(t)) \geq \frac{\lambda_i}{2} |\gamma(t) - x_i^*|^2 \geq \frac{\lambda_i}{2} \pi(\gamma(t))^2$  inside  $B(x_i^*, r_i)$ , and  $s \geq \frac{\lambda_i}{2} r_i^2 > 0$  outside. Since  $\phi$  is increasing,

$$\int \phi(s(\gamma)) |\gamma'| dt \geq \int \phi\left(\frac{\lambda_i}{2} \pi(\gamma)^2\right) |\pi'(\gamma)| dt.$$

The projected path  $u(t) = \pi(\gamma(t))$  starts at  $R$ , ends at  $-R$ , and by the intermediate value theorem visits every value in  $(-R, R)$ . Hence

$$\int \phi\left(\frac{\lambda_i}{2} u^2\right) |du| \geq \int_{-R}^R \phi\left(\frac{\lambda_i}{2} u^2\right) du = 2 \int_0^R \phi\left(\frac{\lambda_i}{2} u^2\right) du,$$

since the integrand is even and nonnegative (backtracking only increases the integral).

*Power-law evaluation.* For  $\phi(s) = s^\alpha$ , the integral evaluates to

$$2 \int_0^R \left(\frac{\lambda_i}{2}\right)^\alpha u^{2\alpha} du = \frac{2(\lambda_i/2)^\alpha}{2\alpha+1} R^{2\alpha+1} = \frac{2(\lambda_i/2)^\alpha}{2\alpha+1} \left(\frac{2\epsilon}{\Lambda_i}\right)^{\alpha+1/2},$$

which is  $\Theta(\epsilon^{\alpha+1/2})$ . Combining with the upper bound from Lemma 6.1 gives  $\text{diam}(X_\epsilon^{(i)}, d_\phi) = \Theta(\epsilon^{\alpha+1/2})$ .  $\square$

## 8. Generalization to Morse–Bott landscapes

In biological applications, a “cell type” is not a single expression state but a manifold of states sharing the same phenotypic identity. We therefore generalize the framework from isolated minima to compact attracting submanifolds.

**Definition 8.1** (Morse–Bott landscape) *We say that  $s : \mathbb{R}^n \rightarrow [0, \infty)$  is a Morse–Bott landscape if the following hold:*

(B1) *The zero set  $\{s = 0\}$  is a disjoint union  $M_1 \sqcup \dots \sqcup M_k$  of compact connected smooth submanifolds of  $\mathbb{R}^n$  (each without boundary).*

(B2) *At every  $p \in M_i$  the normal Hessian*

$$H_p^\perp := \text{Hess } s(p)|_{(T_p M_i)^\perp}$$

*is positive definite.*

(B3) *(Ridge separation) There exists  $c > 0$  such that the connected components of  $\{s < c\}$  containing distinct  $M_i$  have disjoint closures.*

(B4) *(Properness)  $s(x) \rightarrow \infty$  as  $|x| \rightarrow \infty$ .*

Condition (B2) is the standard Morse–Bott nondegeneracy:  $s$  is allowed to be constant (zero) along  $M_i$  but must grow quadratically in the normal directions. By compactness of each  $M_i$ , the normal eigenvalues are uniformly bounded: there exist  $0 < \lambda_i^\perp \leq \Lambda_i^\perp$  such that  $\lambda_i^\perp |\mathbf{v}|^2 \leq \langle H_p^\perp \mathbf{v}, \mathbf{v} \rangle \leq \Lambda_i^\perp |\mathbf{v}|^2$  for all  $p \in M_i$  and  $\mathbf{v} \in (T_p M_i)^\perp$ .

**Lemma 8.2** (Tubular quadratic control) *For each  $i$  there exists  $\rho_i > 0$  such that for all  $x$  with  $\text{dist}(x, M_i) < \rho_i$ ,*

$$\frac{\lambda_i^\perp}{2} \text{dist}(x, M_i)^2 \leq s(x) \leq \frac{\Lambda_i^\perp}{2} \text{dist}(x, M_i)^2.$$

*Proof* In a tubular neighborhood of  $M_i$ , write  $x = \exp_p(\mathbf{v})$  where  $p = \pi_i(x)$  is the nearest-point projection onto  $M_i$  and  $\mathbf{v} \in (T_p M_i)^\perp$  with  $|\mathbf{v}| = \text{dist}(x, M_i)$ . Since  $s|_{M_i} = 0$  and  $\nabla s|_{M_i} = 0$  (because  $M_i \subset \{s = 0\}$  with  $s \geq 0$ ), the second-order Taylor expansion along the fiber gives  $s(x) = \frac{1}{2} \langle H_p^\perp \mathbf{v}, \mathbf{v} \rangle + O(|\mathbf{v}|^3)$ . The uniform eigenvalue bounds and a compactness argument yield the claim for  $\rho_i$  small enough.  $\square$

**Lemma 8.3** (Intra-component collapse in the Morse–Bott setting) *For each  $i$  there exists  $D_i > 0$  such that for all sufficiently small  $\varepsilon$ ,*

$$\text{diam}(X_\varepsilon^{(i)}, d_\phi) \leq D_i \phi(\kappa_i^\perp \varepsilon) \sqrt{\varepsilon},$$

where  $\kappa_i^\perp = \Lambda_i^\perp / \lambda_i^\perp$  and  $X_\varepsilon^{(i)}$  is the component of  $\{s \leq \varepsilon\}$  containing  $M_i$ . In particular,  $\text{diam}(X_\varepsilon^{(i)}, d_\phi) \rightarrow 0$ .

*Proof* Fix  $x, x' \in X_\varepsilon^{(i)}$ . Let  $p = \pi_i(x)$  and  $p' = \pi_i(x')$  be their nearest points on  $M_i$ . Since  $M_i$  is connected, there exists a path  $\sigma : [0, 1] \rightarrow M_i$  from  $p$  to  $p'$ . Along  $\sigma$ ,  $s = 0$ , so  $\phi(s(\sigma)) = 0$  and the  $d_\phi$ -length of  $\sigma$  is zero. Therefore

$$d_\phi(x, x') \leq d_\phi(x, p) + d_\phi(p, p') + d_\phi(p', x') = d_\phi(x, p) + 0 + d_\phi(p', x').$$

For  $d_\phi(x, p)$ : parametrize the normal segment as  $\gamma(t) = p + t(x - p)$  for  $t \in [0, 1]$ . Since  $|\gamma(t) - p| = t \text{dist}(x, M_i)$ , Lemma 8.2 gives  $s(\gamma(t)) \leq \frac{\Lambda_i^\perp}{2} t^2 \text{dist}(x, M_i)^2 \leq \frac{\Lambda_i^\perp}{\lambda_i^\perp} \varepsilon = \kappa_i^\perp \varepsilon$ , where the second inequality uses  $\text{dist}(x, M_i)^2 \leq 2\varepsilon/\lambda_i^\perp$  (from Lemma 8.2). Then  $d_\phi(x, p) \leq \phi(\kappa_i^\perp \varepsilon) \text{dist}(x, M_i) \leq \phi(\kappa_i^\perp \varepsilon) \sqrt{2\varepsilon/\lambda_i^\perp}$ , exactly as in Lemma 4.3. The same bound applies to  $d_\phi(p', x')$ , giving the claim.  $\square$

The inter-basin analysis (Section 5) carries over verbatim: any path from  $M_i$  to  $M_j$  must cross the ridge  $\{s \geq c\}$  and traverse the Euclidean gap, so  $C_{ij} := d_\phi(M_i, M_j) > 0$ .

**Theorem 8.4** (Canalization for Morse–Bott landscapes) *Under (B1)–(B4), with  $d_{\mathcal{F}}(i, j) = C_{ij} = d_\phi(M_i, M_j)$ ,*

$$(X_\varepsilon, d_\phi) \xrightarrow{\text{GH}} (\{1, \dots, k\}, d_{\mathcal{F}}) \quad (\varepsilon \downarrow 0).$$

*Proof* The proof is identical to that of Theorem 6.2: the correspondence  $(x, i) \in \mathcal{R}$  iff  $x \in X_\varepsilon^{(i)}$  has distortion tending to 0 by Lemmas 8.3 and the Morse–Bott analogue of Lemma 5.3.  $\square$

**Remark 8.5** (Biological interpretation) *In the Morse–Bott picture, each cell type  $i$  is modelled by a manifold  $M_i$  of expression states—capturing the well-known fact that cells of the same type exhibit heterogeneous gene expression. Canalization now means that the entire manifold  $M_i$ , together with a tube of nearby states, collapses to a single point in the limit: the internal variability within a cell type is metrically negligible compared to the distance between types.*

## 9. Connection to Freidlin–Wentzell theory

We now connect the warped metric to the stochastic dynamics most commonly associated with Waddington’s landscape: the overdamped Langevin equation

$$dX_t = -\nabla s(X_t) dt + \sqrt{2T} dW_t, \tag{9.1}$$

where  $T > 0$  is a noise intensity (“temperature”) and  $W_t$  is standard Brownian motion in  $\mathbb{R}^n$ .

*The quasi-potential as a warped length metric*

In the Freidlin–Wentzell theory (Freidlin and Wentzell, 1998), the *quasi-potential* from a stable equilibrium  $a$  to a state  $b$  is

$$V(a, b) = \inf_{T > 0} \inf_{\substack{\gamma \in C^1([0, T]; \mathbb{R}^n) \\ \gamma(0) = a, \gamma(T) = b}} \frac{1}{2} \int_0^T |\dot{\gamma}(t) + \nabla s(\gamma(t))|^2 dt.$$

For the gradient drift  $b(x) = -\nabla s(x)$ , the action functional simplifies to a reparametrization-invariant form. The following result is classical (see, e.g., (Freidlin and Wentzell, 1998, Ch. 4) and (Heymann and Vanden-Eijnden, 2008)); we include the short proof for the reader’s convenience and to set notation.

**Proposition 9.1** (Geometric action (Freidlin and Wentzell 1998; Heymann and Vanden-Eijnden 2008))  
*For two global minima  $x_i^*, x_j^*$  with  $s(x_i^*) = s(x_j^*) = 0$ ,*

$$V(x_i^*, x_j^*) = \inf_{\gamma} \int |\nabla s(\gamma)| |d\gamma| =: d_{|\nabla s|}(x_i^*, x_j^*), \quad (9.2)$$

where the infimum is over piecewise  $C^1$  paths connecting  $x_i^*$  to  $x_j^*$ .

*Proof* Expanding the action:

$$\frac{1}{2} \int_0^T |\dot{\gamma} + \nabla s(\gamma)|^2 dt = \frac{1}{2} \int_0^T |\dot{\gamma}|^2 dt + \int_0^T \langle \dot{\gamma}, \nabla s(\gamma) \rangle dt + \frac{1}{2} \int_0^T |\nabla s(\gamma)|^2 dt.$$

The cross term equals  $s(\gamma(T)) - s(\gamma(0)) = 0$  (since both endpoints have  $s = 0$ ), so the action becomes  $\frac{1}{2} \int |\dot{\gamma}|^2 dt + \frac{1}{2} \int |\nabla s(\gamma)|^2 dt$ . By the AM–GM inequality,  $\frac{1}{2}(|\dot{\gamma}|^2 + |\nabla s|^2) \geq |\dot{\gamma}| |\nabla s|$ , with equality if and only if  $|\dot{\gamma}(t)| = |\nabla s(\gamma(t))|$  for all  $t$ . The right-hand side  $\int |\dot{\gamma}| |\nabla s(\gamma)| dt = \int |\nabla s(\gamma)| |d\gamma|$  is reparametrization-invariant, giving (9.2). Equality is achieved along *instantons*: paths satisfying  $\dot{\gamma} = +\nabla s(\gamma)$  (time-reversed gradient flow) reparametrized to match the arclength condition.  $\square$

**Remark 9.2** (Position-dependent vs. scalar weight) *The quasi-potential defines a warped length metric  $d_{|\nabla s|}$  with position-dependent weight  $\psi(x) = |\nabla s(x)|$ , which in general is not a function of  $s(x)$  alone. Our framework uses  $\phi(s(x))$ , a weight depending on  $x$  only through the scalar landscape value. The two frameworks coincide when level sets of  $s$  are round (all eigenvalues of Hess  $s$  equal); in general they are related as follows.*

**Proposition 9.3** (Comparison near minima) *In the quadratic regime  $|x - x_i^*| < r_i$ , the gradient satisfies*

$$\frac{\sqrt{2}\lambda_i}{\sqrt{\Lambda_i}} \sqrt{s(x)} \leq |\nabla s(x)| \leq \frac{\sqrt{2}\Lambda_i}{\sqrt{\lambda_i}} \sqrt{s(x)}.$$

Consequently, for paths  $\gamma$  contained in  $B(x_i^*, r_i)$ ,

$$\frac{\sqrt{2}\lambda_i}{\sqrt{\Lambda_i}} \ell_{\sqrt{s}}(\gamma) \leq \ell_{|\nabla s|}(\gamma) \leq \frac{\sqrt{2}\Lambda_i}{\sqrt{\lambda_i}} \ell_{\sqrt{s}}(\gamma), \quad (9.3)$$

where  $\ell_w(\gamma) = \int w(\gamma) |d\gamma|$  denotes the weighted length. When  $\lambda_i = \Lambda_i$  (isotropic minimum) both constants equal  $\sqrt{2\lambda_i}$ .

*Proof* Write  $h = x - x_i^*$  and  $H_i = \text{Hess } s(x_i^*)$ . In the quadratic regime,  $\nabla s(x) = H_i h + O(|h|^2)$  and  $s(x) = \frac{1}{2} h^\top H_i h + O(|h|^3)$ . The eigenvalue bounds on  $H_i$  give  $\lambda_i |h| \leq |H_i h| \leq \Lambda_i |h|$  and  $\frac{\lambda_i}{2} |h|^2 \leq s(x) \leq \frac{\Lambda_i}{2} |h|^2$ . Therefore  $|h| \leq \sqrt{2s(x)/\lambda_i}$  and  $|h| \geq \sqrt{2s(x)/\Lambda_i}$ , so

$$|\nabla s(x)| \geq \lambda_i |h| \geq \lambda_i \sqrt{2s(x)/\Lambda_i} = \frac{\sqrt{2}\lambda_i}{\sqrt{\Lambda_i}} \sqrt{s(x)}$$

and

$$|\nabla s(x)| \leq \Lambda_i |h| \leq \Lambda_i \sqrt{2s(x)/\lambda_i} = \frac{\sqrt{2}\Lambda_i}{\sqrt{\lambda_i}} \sqrt{s(x)}.$$

Integrating along  $\gamma$  yields (9.3).  $\square$

**Remark 9.4** (The dynamically natural choice  $\alpha = \frac{1}{2}$ ) *Proposition 9.3 shows that near minima the quasi-potential weight  $|\nabla s|$  is comparable to  $\sqrt{s}$ , which corresponds to  $\phi(s) = s^{1/2}$  in our framework. The choice  $\alpha = \frac{1}{2}$  therefore has a distinguished dynamical meaning: it is the unique power-law exponent for which the warped metric locally approximates the Freidlin–Wentzell quasi-potential. With this choice the collapse rate is  $\text{diam} = \Theta(\varepsilon)$  (linear in  $\varepsilon$ ), and the inter-basin distances  $C_{ij}$  approximate the quasi-potential barriers controlling rare transitions between cell types.*

### Explicit comparison in one dimension

For  $n = 1$  with  $s(x) = (x^2 - 1)^2$ , the quasi-potential between the two minima is

$$V(-1, 1) = \int_{-1}^1 |s'(x)| \, dx = \int_{-1}^1 4|x|(1-x^2) \, dx = 2.$$

This is consistent with the classical formula for gradient systems:  $V(x_i^*, x_j^*) = 2s(x_s)$  where  $x_s$  is the lowest saddle separating the two minima (Freidlin and Wentzell, 1998, Ch. 4). Here  $s(0) = 1$  and  $V = 2 \cdot 1 = 2$ . The warped length distance with  $\phi(s) = \sqrt{s}$  is

$$d_{\sqrt{s}}(-1, 1) = \int_{-1}^1 (1-x^2) \, dx = \frac{4}{3}.$$

The ratio  $V/d_{\sqrt{s}} = 3/2$  reflects the anisotropy of  $|s'(x)|/\sqrt{s(x)} = 4|x|$ , which ranges from 0 at the saddle to 4 at the minima.

## 10. Examples

### 10.1. A one-dimensional double well

Let  $n = 1$  and  $s(x) = (x^2 - 1)^2$ , which has two nondegenerate global minima at  $x^* = \pm 1$ . Take  $\phi(s) = s^\alpha$  with  $\alpha > 0$ . For  $\varepsilon$  small,  $X_\varepsilon$  is the union of two intervals around  $\pm 1$  and the limit metric space has two points.

In one dimension the minimizing curve between minima is the interval path, hence

$$C_{12} = d_\phi(-1, 1) = \int_{-1}^1 \phi(s(x)) \, dx = \int_{-1}^1 (x^2 - 1)^{2\alpha} \, dx < \infty.$$

By Proposition 7.2,  $\text{diam}(X_\varepsilon^{(i)}, d_\phi) = \Theta(\varepsilon^{\alpha+1/2})$ . For the dynamically natural choice  $\alpha = \frac{1}{2}$  (Remark 9.4), the collapse is linear:  $\text{diam} = \Theta(\varepsilon)$ .

### 10.2. A two-dimensional toggle switch

A genetic toggle switch consists of two genes with mutual repression: when gene 1 is highly expressed, it suppresses gene 2, and vice versa (Gardner et al., 2000). This produces bistability with two stable cell fates. The standard ODE model uses Hill-function kinetics and is not a gradient system; its quasi-potential must be computed numerically (Wang et al., 2011). Here we use an explicit polynomial proxy that captures the essential bistable topology—two minima separated by a saddle—while remaining analytically tractable as a Morse function.

We model this via the landscape

$$s(x, y) = [(x - 1)^2 + y^2][x^2 + (y - 1)^2], \quad (10.1)$$

on  $\mathbb{R}^2$ , where  $x$  and  $y$  represent the expression levels of genes 1 and 2 respectively.

**Zero set and minima.** The first factor vanishes only at  $p_1 = (1, 0)$  (gene 1 on, gene 2 off) and the second only at  $p_2 = (0, 1)$  (gene 2 on, gene 1 off), so  $\{s = 0\} = \{p_1, p_2\}$ .

**Hessian computation.** Write  $s = fg$  with  $f = (x - 1)^2 + y^2$  and  $g = x^2 + (y - 1)^2$ . At  $p_1 = (1, 0)$ :  $f = 0$ ,  $g = 2$ ,  $\nabla f = (0, 0)$ , so  $\text{Hess } s|_{p_1} = g(p_1) \cdot \text{Hess } f|_{p_1} = 2 \cdot 2I = 4I$ . By the symmetry  $(x, y) \mapsto (y, x)$  that exchanges  $p_1 \leftrightarrow p_2$ , also  $\text{Hess } s|_{p_2} = 4I$ . Both minima are isotropic with  $\lambda_i = \Lambda_i = 4$  and condition number  $\kappa_i = 1$ .

**Saddle point.** The unique critical point in the interior region  $\{f > 0, g > 0\}$  is the midpoint  $p_s = (\frac{1}{2}, \frac{1}{2})$  (see the completeness check below); it has  $s(p_s) = \frac{1}{4}$ . The Hessian at  $p_s$  has eigenvalues  $\pm 2$ , confirming a nondegenerate saddle. There are no other critical points (see below), so  $s$  is Morse with exactly two minima and one saddle.

**Verification of (A1)–(A4).** (A1) Morse: checked above. (A2) Two global minima at  $p_1, p_2$  with  $s = 0$ . (A3) Ridge: for any  $c < \frac{1}{4}$ , the components of  $\{s < c\}$  around  $p_1$  and  $p_2$  are separated by the saddle region. (A4) Properness:  $s(x, y) \rightarrow \infty$  as  $|(x, y)| \rightarrow \infty$  since  $s$  is a degree-4 polynomial with positive leading terms.

**Completeness of critical-point analysis.** Write  $s = fg$  with  $f = (x - 1)^2 + y^2$  and  $g = x^2 + (y - 1)^2$ , so  $f, g \geq 0$  and

$$\nabla s = g\nabla f + f\nabla g.$$

If  $f = 0$  then  $(x, y) = (1, 0) = p_1$ , and if  $g = 0$  then  $(x, y) = (0, 1) = p_2$ . Otherwise  $f, g > 0$  and the critical-point equations become

$$g(x - 1) + fx = 0, \quad gy + f(y - 1) = 0.$$

Introduce centered coordinates  $u = x - \frac{1}{2}$ ,  $v = y - \frac{1}{2}$ , set  $r^2 = u^2 + v^2$ , and note

$$f = r^2 + (v - u) + \frac{1}{2}, \quad g = r^2 + (u - v) + \frac{1}{2}.$$

With  $a := r^2 + \frac{1}{2}$  and  $d := v - u$  (so  $f = a + d$  and  $g = a - d$ ), the two equations simplify to

$$2au + d = 0, \quad 2av - d = 0.$$

Hence  $v = -u$ , and then  $d = v - u = -2u$  together with  $d = -2au$  gives either  $u = 0$  (so  $v = 0$  and  $(x, y) = (\frac{1}{2}, \frac{1}{2}) = p_s$ ) or  $a = 1$  (so  $r^2 = \frac{1}{2}$  and  $(u, v) = (\pm\frac{1}{2}, \mp\frac{1}{2})$ , which correspond to  $p_1$  and  $p_2$ ). Therefore  $\{p_1, p_2, p_s\}$  are all critical points of  $s$ .

**Collapse rate.** Since  $\kappa_i = 1$  and  $\lambda_i = 4$ , Proposition 7.2 and Lemma 6.1 give

$$\text{diam}(X_\varepsilon^{(i)}, d_\phi) = \Theta(\varepsilon^{\alpha+1/2}) \quad (i = 1, 2).$$

For  $\alpha = \frac{1}{2}$ :  $\text{diam} = \Theta(\varepsilon)$ .

**Inter-basin distance.** The straight-line path  $\gamma(t) = (1-t, t)$ ,  $t \in [0, 1]$ , connects  $p_1$  to  $p_2$  with  $|\gamma'| = \sqrt{2}$  and

$$s(\gamma(t)) = 4t^2(1-t)^2.$$

This gives the upper bound

$$C_{12} \leq \sqrt{2} \int_0^1 \phi(4t^2(1-t)^2) dt.$$

For  $\phi(s) = s^\alpha$ :

$$C_{12} \leq \sqrt{2} \cdot 4^\alpha \int_0^1 t^{2\alpha} (1-t)^{2\alpha} dt = \sqrt{2} \cdot 4^\alpha B(2\alpha+1, 2\alpha+1),$$

where  $B$  is the Beta function. Some explicit values:

$\alpha$	1/2	1	2
$C_{12} \leq$	$\frac{\sqrt{2}}{3} \approx 0.471$	$\frac{2\sqrt{2}}{15} \approx 0.189$	$\frac{8\sqrt{2}}{315} \approx 0.036$

For a lower bound, any path from  $p_1$  to  $p_2$  must cross the ridge  $\{s \geq c\}$  (for  $c < \frac{1}{4}$ ) and traverse the Euclidean gap  $w_{12} > 0$  between the basin closures. The general bound  $C_{12} \geq \phi(c) w_{12}$  from Lemma 5.2 applies.

**Biological interpretation.** The landscape (10.1) models the two cell fates arising from mutual repression: state  $p_1 = (1, 0)$  is the “gene-1-dominant” fate and  $p_2 = (0, 1)$  is the “gene-2-dominant” fate. The warped metric makes each cell type metrically small (collapse rate  $\Theta(\varepsilon^{\alpha+1/2})$ ) while preserving the inter-fate distance  $C_{12}$ . The ratio  $C_{12}/\text{diam}$  quantifies the *canalization strength*: larger values indicate more robust separation between fates. In the dynamically natural scaling  $\alpha = \frac{1}{2}$ , the inter-basin distance  $C_{12} \approx 0.471$  should be compared with the Freidlin–Wentzell quasi-potential  $\tilde{V}(p_1, p_2)$ , which governs the rate of noise-induced fate switching.

## 11. Discussion and outlook

We have given a rigorous geometric formalization of canalization—the robustness of developmental outcomes—via Gromov–Hausdorff collapse of conformally warped sublevel sets. The main theorem (Theorem 6.2) and its Morse–Bott extension (Theorem 8.4) show that low sublevel sets of a landscape  $s$  collapse to a finite discrete space under the warped metric  $d_\phi$ . The sharp collapse rate  $\Theta(\varepsilon^{\alpha+1/2})$  (Proposition 7.2) quantifies how rapidly intra-basin distances vanish relative to inter-basin distances. The Freidlin–Wentzell connection (Propositions 9.1 and 9.3) anchors the choice  $\alpha = \frac{1}{2}$  in stochastic dynamics.

Several natural extensions remain:

- (E1) *Graph limits:* replacing  $X_\varepsilon$  by neighborhoods of the gradient-flow skeleton (minima, index-1 saddles, connecting orbits) may produce metric-graph GH limits, connecting to the Reeb graph and persistent homology.

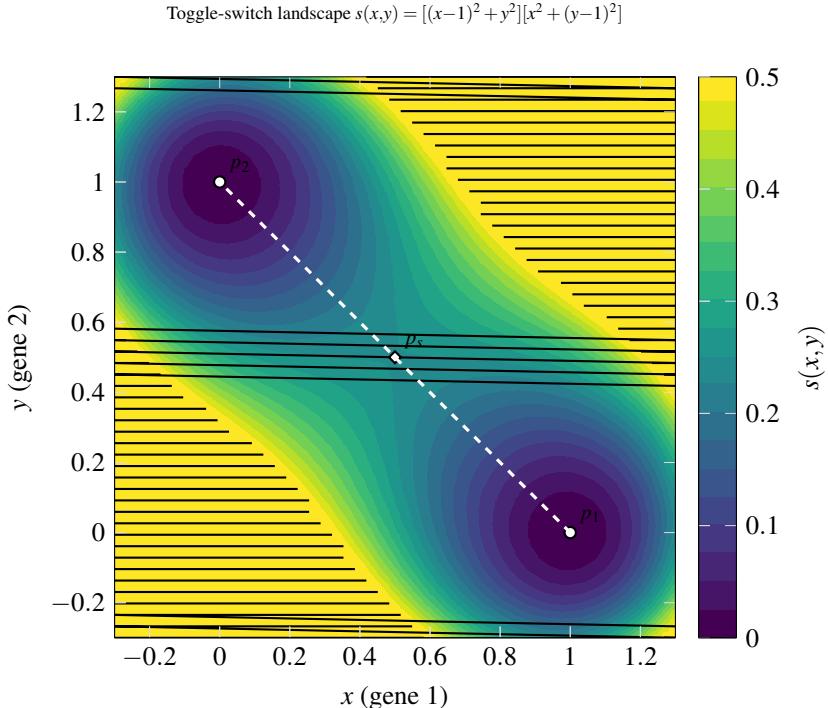


Figure 2. Contour plot of the toggle-switch landscape (10.1). The two minima  $p_1 = (1,0)$  and  $p_2 = (0,1)$  (circles) represent the two cell fates; the saddle  $p_s = (\frac{1}{2}, \frac{1}{2})$  (diamond) separates the basins. The dashed line shows the straight-line path used to compute the upper bound on  $C_{12}$ . For small  $\varepsilon$ , the sublevel set  $\{s \leq \varepsilon\}$  consists of two small discs around  $p_1$  and  $p_2$  that collapse under  $d_\phi$ .

(E2) *Curvature analysis:* the sectional curvature of  $g_\phi$  diverges near  $\{s = 0\}$ ; characterizing the blow-up rate and establishing Alexandrov curvature bounds  $\kappa(\varepsilon)$  on the sublevel sets would place the paper in the RCD literature.

(E3) *Data-driven  $\phi$ :* different near-zero behaviors of  $\phi$  correspond to different canalization strengths. Given single-cell RNA-seq data and an inferred landscape  $\hat{s}$ , the warping function  $\phi$  could in principle be estimated from observed variability scaling within basins.

(E4) *Non-gradient dynamics:* many gene-regulatory ODEs are not gradient systems. Extending the framework to quasi-potential landscapes defined via paths of minimum Freidlin–Wentzell action would cover this biologically important case.

(E5) *Wasserstein collapse:* placing the Boltzmann measure  $\mu_T \propto e^{-s/T}$  on  $(X_\varepsilon, d_\phi)$  and proving Wasserstein convergence to a discrete measure on the minima would connect to optimal transport.

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