



Integrated Harnack Inequalities

Joint with Masha Gordina

Bruce Driver

Visiting Miller Professor at UC – Berkeley and

Department of Mathematics, 0112

University of California at San Diego, USA

<http://math.ucsd.edu/~driver>

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Example; $G = \mathbb{R}^n$

Example 1. Suppose $A \in \mathbb{R}^n$,

$$p_T(x) = \left(\frac{1}{2\pi T} \right)^{n/2} e^{-\frac{|x|^2}{2T}} \quad \forall x \in \mathbb{R}^n, \text{ and}$$

$$W_A(x) = -(\partial_A \ln p_T)(x) = \frac{x \cdot A}{T}.$$

Then

$$\int_{\mathbb{R}^n} e^{W_A(x)} p_T(x) dx = e^{\frac{|A|^2}{2T}},$$

and for all $y \in \mathbb{R}^n$,

$$\left(\int_{\mathbb{R}^n} \left[\frac{p_T(x-y)}{p_T(x)} \right]^p p_T(x) dx \right)^{1/p} = \exp \left(\frac{(p-1)}{2T} d^2(e, y) \right). \quad (1)$$

Proof: (Proof of Eq. (1).)

$$\begin{aligned} \int_{\mathbb{R}^n} \left[\frac{p_T(x-y)}{p_T(x)} \right]^p p_T(x) dx &= \int_{\mathbb{R}^n} \exp \left(\frac{p}{T} x \cdot y - \frac{p}{2T} |y|^2 \right) p_T(x) dx \\ &= \exp \left(\frac{p^2 - p}{2T} |y|^2 \right). \end{aligned}$$

Q.E.D.

Application (Cameron-Martin Theorem)

Theorem 1. Let (W, H, μ) be an abstract Wiener space and $h \in H$. Then there exists $Z_h \in \cap_{p' < \infty} L^{p'}(\mu)$ such that, for all measurable, $f : W \rightarrow [0, \infty)$,

$$\int_W f(h + w) d\mu(w) = \int_W f(w) Z_h(w) d\mu(w)$$

and

$$\|Z_h\|_{p'} \leq \exp\left(\frac{p' - 1}{2} \|h\|_H^2\right).$$

Proof:

- Let $f : W \rightarrow [0, \infty)$ be a bounded and continuous function.
- Let $\{e_j\}_{j=1}^{\infty} \subset H_* \cong W^*$ be an orthonormal basis for H , and

$$P_n(w) = \sum_{j=1}^n \langle w, e_j \rangle_H e_j \text{ with } e_j \in H_*.$$

Then;

$$\begin{aligned}
\int_W f \circ P_n(h + w) d\mu(w) &= \int_{P_n H} f(P_n h + x) p_1^{P_n H}(x) dx \\
&= \int_{P_n H} f(x) p_1^{P_n H}(x - P_n h) dx \\
&= \int_{P_n H} f(x) \frac{p_1^{P_n H}(x - P_n h)}{p_1^{P_n H}(x)} p_1^{P_n H}(x) dx \\
&\leq \|f\|_{L^p(p_1^{P_n H}(x) dx)} \cdot \left\| \frac{p_1^{P_n H}(\cdot - P_n h)}{p_1^{P_n H}(\cdot)} \right\|_{L^{p'}(p_1^{P_n H}(x) dx)} \\
&\leq \|f\|_{L^p(p_1^{P_n H}(x) dx)} \exp\left(\frac{p' - 1}{2} \|P_n h\|_H^2\right).
\end{aligned}$$

Now let $n \rightarrow \infty$ to find

$$\int_W f(h + w) d\mu(w) \leq \|f\|_{L^p(\mu)} \exp\left(\frac{p' - 1}{2} \|h\|_H^2\right).$$

Since, $L^{p^*} \cong L^{p'}$, there exists $Z_h \in L^{p'}(\mu)$ such that

$$\int_W f(h+w) d\mu(w) = \int_W f(w) Z_h(w) d\mu(w)$$

and

$$\|Z_h\|_{p'} \leq \exp\left(\frac{p'-1}{2} \|h\|_H^2\right).$$

Q.E.D.

Setup for Extending to Lie Groups

- G = a connected finite dimensional uni-modular Lie group.
- $\mathfrak{g} = \text{Lie}(G)$
- $(\cdot, \cdot) = (\cdot, \cdot)_{\mathfrak{g}}$ – an inner product on \mathfrak{g} , $|A|_{\mathfrak{g}} := \sqrt{(A, A)}$ for all $A \in \mathfrak{g}$.
- Use $(\cdot, \cdot)_{\mathfrak{g}}$ to make a left invariant Riemannian metric on G .
- $d(x, y)$ is the associated Riemannian distance function
- For $A \in \mathfrak{g}$ let \tilde{A} denote the unique left invariant vector field on G such that $\tilde{A}(e) = A \in \mathfrak{g}$.
- $\Delta = \sum_{i=1}^{\dim \mathfrak{g}} \tilde{A}_i^2$ where $\{A_i\}_{i=1}^{\dim \mathfrak{g}}$ is an orthonormal basis for \mathfrak{g} .

Example

Let $G = \mathbb{R}^3$ with the multiplication law,

$$(x, y, z) (x', y', z') := \left(x + x', y + y', z + z' + \frac{1}{2} (xy' - yx') \right).$$

- A basis of left invariant vector fields is;

$$\tilde{A}_1 = \frac{\partial}{\partial x} - \frac{1}{2} y \frac{\partial}{\partial z},$$

$$\tilde{A}_2 = \frac{\partial}{\partial y} + \frac{1}{2} x \frac{\partial}{\partial z}$$

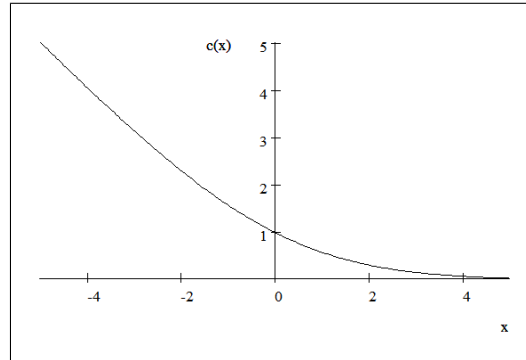
$$\tilde{A}_3 = [\tilde{A}_1, \tilde{A}_2] = \frac{\partial}{\partial z}$$

- The Laplacian is then

$$\begin{aligned} \Delta &= \tilde{A}_1^2 + \tilde{A}_2^2 + \tilde{A}_3^2 \\ &= \frac{\partial^2}{\partial x^2} + \frac{\partial^2}{\partial y^2} + \left(x \frac{\partial}{\partial y} - y \frac{\partial}{\partial x} \right) \frac{\partial}{\partial z} + \left(\frac{x^2 + y^2}{4} + 1 \right) \frac{\partial^2}{\partial z^2} \end{aligned}$$

Definition 2. Let $c(0) = 1$ and for $x \neq 0$,

$$c(x) = \frac{x}{e^x - 1}. \quad (2)$$



Theorem 3.1 (*Heat Kernel Jacobian Estimate*. Also see [Driver, 1997b]). Let $T > 0$,

$$p_T(x) = \left(e^{T\Delta/2} \delta_e \right) (x),$$

and $\text{Ric} \geq kId$. Then for every $y \in G$ and $p \in [1, \infty)$,

$$\left(\int_G \left[\frac{p_T(xy^{-1})}{p_T(x)} \right]^p p_T(x) dx \right)^{1/p} \leq \exp \left(\frac{c(kT)(p-1)}{2T} d^2(e, y) \right). \quad (3)$$

Definition 3.2. For $A \in \mathfrak{g}$, let

$$W_A(x) := -(\tilde{A} \ln p_T)(x)$$

or equivalently,

$$\int_G \tilde{A} f(x) p_T(x) dx = \int_G f(x) W_A(x) p_T(x) dx \text{ for all } f \in C_c^\infty(G).$$

The proof of Theorem 3.1 relies on the following key exponential integrability estimate for W_A .

Theorem 3.3 (*Exponential Integrability*). Continuing the notation in Theorem 3.1 and in particular let $c_T(k)$ be as in Eq. (2). Then for every $A \in \mathfrak{g}$,

$$\int_G e^{W_A(x)} p_T(x) dx \leq \exp\left(\frac{c(kT)}{2T} |A|_{\mathfrak{g}}^2\right).$$

(4)

Applications

Let $d\nu_T(x) = p_T(x) dx$ be the heat kernel measure.

1. *Harmonic Pointwise Upper Bounds:* If $f \in L^p(\nu_T)$ and f is harmonic, then $P_T f = f$ and

$$|f(y)| \leq \|f\|_{L^p(\nu)} \exp\left(\frac{c(kT)}{2T(p-1)} d^2(e, y)\right). \quad (5)$$

2. *Holomorphic Pointwise Upper Bounds:* Eq. (5) holds if f is holomorphic but the constant is certainly not optimal. If $p = 2$, one can replace $\frac{c(kT)}{2T}$ by $\frac{1}{4T}$.

3. *Quasi-Invariance in Infinite Dimensions:*

- (a) Certain Projective Groups like Loop and Path groups, see for example, [Driver, 1997c] and [Fang, 1999]. Also see [Airault & Malliavin, 2006, JFA 241].
- (b) Certain infinite dimensional Nilpotent groups like an infinite dimensional Heisenberg groups, see Masha Gordina's talk.

Derivative Formulas

Theorem 5.1 (*Divergence Representation Formula I*). Assume the Ricci curvature, Ric , on M is bounded. Let $T > 0$ and $\tilde{\ell}$ be a continuous real-valued process $\tilde{\ell}_0 = 0$, $\tilde{\ell}_T = 1$, and there exists a non-random constant, $C < \infty$, such that

$$\int_0^T \left| \frac{d}{d\tau} \tilde{\ell}_\tau \right| d\tau \leq C. \quad (6)$$

Then for every C^1 – vector field, Y , on M with compact support the following identity holds

$$\mathbb{E} [\nabla \cdot Y (\Sigma_T^x(x))] = \mathbb{E} \left[\left\langle Y (\Sigma_T^x(x)), //_T Q_T \int_0^T \tilde{\ell}'_t Q_t^{-1} db_t \right\rangle \right], \quad (7)$$

where $\nabla \cdot Y$ denote the divergence of Y . Here, $\Sigma_t^x \in M$ is Brownian motion, $//_t$ is parallel translation, and Q solves

$$\dot{Q}_t = -\frac{1}{2} \text{Ric} //_t Q_t \quad \text{with } Q_0 = id_{T_x M}. \quad (8)$$

Proof: See [Driver & Thalmaier, 2001].

Q.E.D.

Similar manipulations to those in [Driver, 1997b], shows Theorem 5.1 may be rephrased, in the Lie group setting ($M = G$) as;

Theorem 5.2 (*Divergence Representation Formula II*). Let $A \in \mathfrak{g}$, $\ell \in C^1([0, T], \mathbb{R})$ with $\ell(0) = 0$ and $\ell(T) = 1$ and define;

$$\beta_t := \int_0^t L_{\Sigma_\tau^{-1}*} \delta \Sigma_\tau.$$

Then

$$\boxed{-\left(\tilde{A} \ln p_T\right)(x) = W_A(x) = \mathbb{E} \left[\left(A, \int_0^T \dot{\ell}(\tau) V_\tau d\overleftarrow{\beta}_\tau \right) \middle| \Sigma_T = x \right],} \quad (9)$$

where $\int_0^T \dot{\ell}(\tau) V_\tau d\overleftarrow{\beta}_\tau$ is a backwards Itô integral and V_t satisfies

$$dV_t = \frac{1}{2} V_t \text{Ric}_e dt + V_t D_{\text{od}\beta_t} \text{ with } V_T = Id.$$

Here,

$$D_A : \mathfrak{g} \rightarrow \mathfrak{g} \text{ for all } A \in \mathfrak{g}$$

is the Levi-Civita connection as seen on \mathfrak{g} and $\Sigma_t \in G$ is Brownian motion starting at $e \in G$.

Lemma 5.3. Suppose $k \in \mathbb{R}$ is chosen so that $\text{Ric} \geq kI$, then

$$|V_t^* A|^2 \leq |A|^2 e^{-k(T-t)} \text{ for all } A \in \mathfrak{g}. \quad (10)$$

Proof of Theorem 3.3

Theorem 6.1 (Theorem 3.3).

$$\int_G e^{W_A(x)} p_T(x) dx \leq \exp\left(\frac{c(kT)}{2T} |A|_{\mathfrak{g}}^2\right) \text{ for all } A \in \mathfrak{g}.$$

Proof: Let $\ell \in C^1([0, T], \mathbb{R})$ such that $\ell(0) = 0$ and $\ell(T) = 1$. From Theorem 5.2, Lemma 5.3, Jensen's inequality for conditional expectations, and a standard martingale argument (see the proof of Lemma 7.6 and especially Eq. 7.17 in [Driver, 1997a]), starting with

$$W_A(x) = \mathbb{E} \left[\left(A, \int_0^T \dot{\ell}(\tau) V_\tau d\overleftarrow{\beta}_\tau \right) \middle| \Sigma_T = x \right],$$

we have...

$$\begin{aligned}
\int_G e^{W_A} d\nu_T &= \mathbb{E} \left[\exp \left(\mathbb{E} \left[\left(A, \int_0^T \dot{\ell}(\tau) V_\tau d\overleftarrow{\beta}_\tau \right) \middle| \sigma(\Sigma_T) \right] \right) \right] \\
\text{(Jensen)} &\leq \mathbb{E} \left[\mathbb{E} \left[\exp \left(\left(A, \int_0^T \dot{\ell}(\tau) V_\tau d\overleftarrow{\beta}_\tau \right) \right) \middle| \sigma(\Sigma_T) \right] \right] \\
&\leq \mathbb{E} \left[\exp \left(\left(A, \int_0^T \dot{\ell}(\tau) V_\tau d\overleftarrow{\beta}_\tau \right) \right) \right] \\
\text{(Mrtg. Est.)} &\leq \exp \left(\frac{1}{2} \left\| \int_0^T \dot{\ell}^2(\tau) |V_\tau^* A|^2 d\tau \right\|_{L^\infty(P)} \right) \\
\text{(Lemma 5.3)} &\leq \exp \left(\frac{|A|^2}{2} \int_0^T \dot{\ell}^2(\tau) e^{-k(T-\tau)} d\tau \right).
\end{aligned}$$

The last expression is minimized by taking,

$$\dot{\ell}(\tau) = \frac{e^{-k\tau} - 1}{e^{-kT} - 1}$$

and this gives;

$$\int_G e^{W_A} d\nu_T \leq \exp \left(\frac{1}{2T} c(kT) |A|^2 \right).$$

Q.E.D.

Exponential Integrability \rightarrow Jacobian Estimate

Our next goal is to show that

Theorem 7.1 (Theorem 3.3).

$$\int_G e^{W_A(x)} p_T(x) dx \leq \exp\left(\frac{c(kT)}{2T} |A|_{\mathfrak{g}}^2\right) \text{ for all } A \in \mathfrak{g}.$$

\Downarrow

Theorem 7.2 (*Heat Kernel Jacobian Estimate*). Let $T > 0$,

$$p_T(x) = \left(e^{T\Delta/2} \delta_e\right)(x),$$

and $\text{Ric} \geq kId$. Then for every $y \in G$ and $p \in [1, \infty)$,

$$\left(\int_G \left[\frac{p_T(xy^{-1})}{p_T(x)}\right]^p p_T(x) dx\right)^{1/p} \leq \exp\left(\frac{c(kT)(p-1)}{2T} d^2(e, y)\right).$$

Some key ideas;

- We may find a path, $A_t \in \mathfrak{g}$ such that $S_1(x) = R_y(x) = xy$ where S_t solves the flow equation,

$$\frac{d}{dt} S_t(x) = (\tilde{A}_t S_t)(x) \text{ with } S_0(x) = x \forall x \in G.$$

- $d(x, y) = \inf \left\{ \int_0^1 |A_t|_{\mathfrak{g}} dt : \text{over all such } t \rightarrow A_t \right\}.$

- If we let $d\mu(x) := p_T(x) dx$, then for $y \in G$,

$$J_1(x) := \frac{d\mu \circ S_1^{-1}}{d\mu}(x) = \frac{p_T(xy^{-1})}{p_T(x)}.$$

Jacobian Estimates Generalities

- Let M be a finite dimensional manifold,
- μ be a smooth positive probability measure on M
- $\|f\|_r := \left(\int_M |f|^r d\mu\right)^{1/r}$
- X_t be a time dependent vector field on M
- S_t denote the flow, i.e. $S_t(m)$ solves,

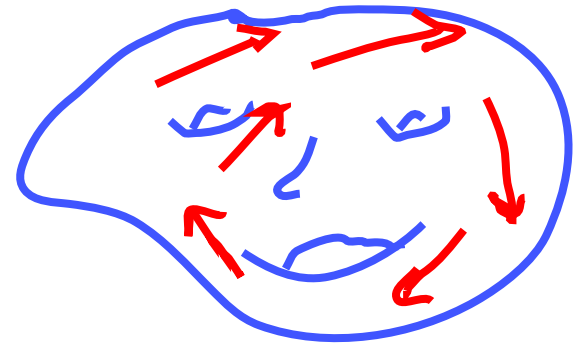
$$\frac{d}{dt}S_t(m) = X_t \circ S_t(m) \text{ with } S_0(m) = m \text{ for all } m \in M. \quad (11)$$

- Assume that X_t is forward complete.
- Let W_t be the μ – **divergence** of X_t , i.e.

$$\int_M X_t f d\mu = \int_M W_t f d\mu \text{ for all } f \in C_c^\infty(M).$$

- Define

$$\mu_t = (S_t)_* \mu = \mu \circ S_t^{-1} \text{ and } J_t = d\mu_t/d\mu.$$



The Galaz-Fontes, Gross, and Sontz Method

Notation 9.1. For any function, $W : M \rightarrow \mathbb{R}$,

$$\mathcal{B}(W) := \ln(\mu(e^W)) = \ln\left(\int_M e^W d\mu\right).$$

- The following is an extension of a theorem of Galaz-Fontes, Gross, and Sontz which extends the pioneering work of Cruzeiro in this direction.

Theorem 9.2 (*Jacobian Estimate* [Galaz-Fontes *et al.*, 2001]). Let $p > 1$ and $r \in C([0, \tau], [1, \infty)) \cap C^1((0, \tau), (1, \infty))$ such that $r(0) = 1$, $r(\tau) = p$ and $\dot{r}(t) > 0$ for $0 < t < \tau$, then

$$\|J_\tau\|_{p'} \leq e^{\Lambda(r)}, \quad (12)$$

where $p' := p/(p-1)$ is the conjugate exponent to p and

$$\Lambda(r) = \Lambda_X(r) := \int_0^\tau \frac{\dot{r}(t)}{r^2(t)} \mathcal{B}\left(\frac{r(t)}{\dot{r}(t)} W_t\right) dt. \quad (13)$$

- We will now use this theorem with

$$M = G \text{ and } d\mu(x) = p_T(x) dx.$$

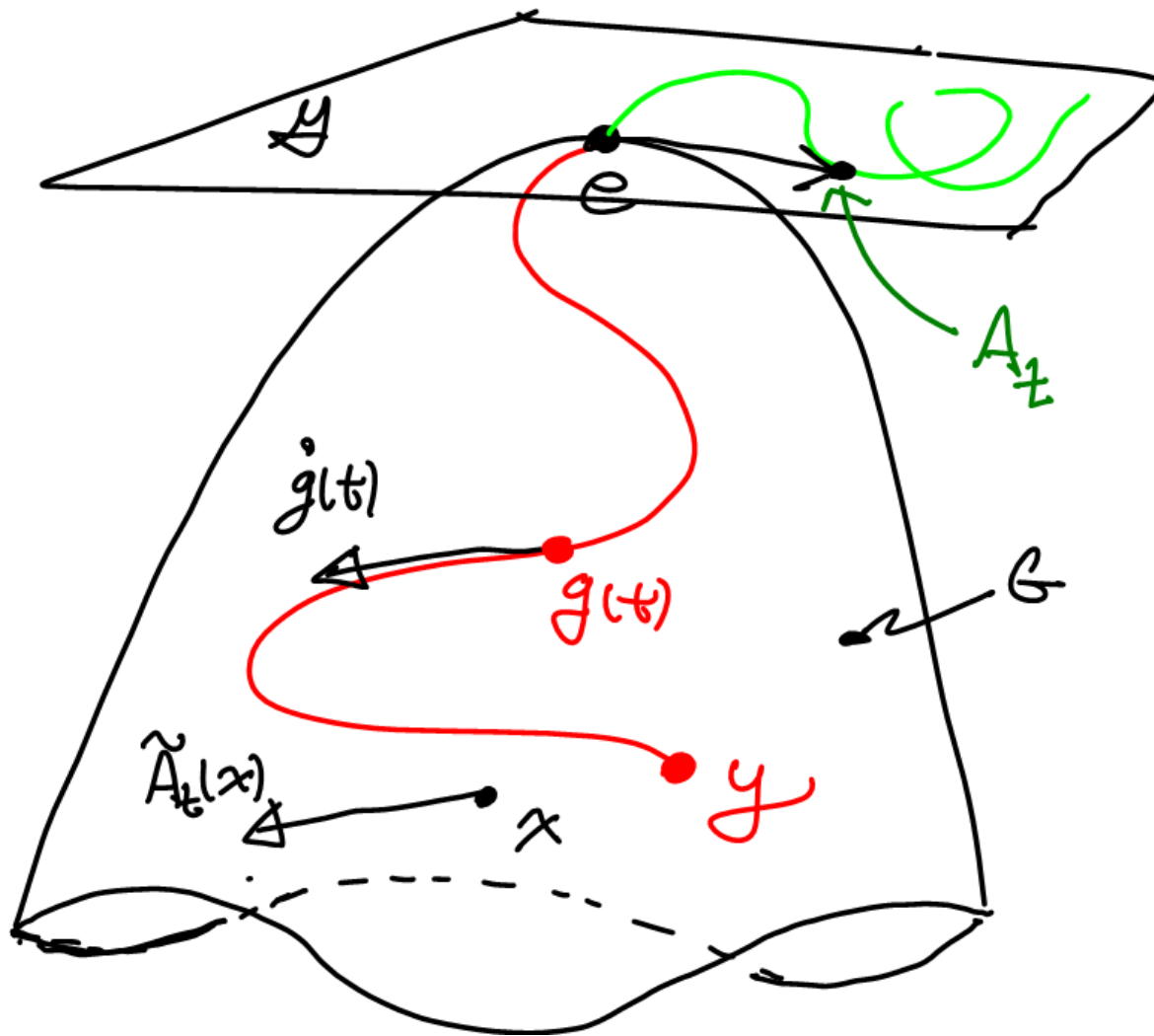
Proof of Theorem 3.1

Theorem 10.1 (Theorem 3.1). If $\text{Ric} \geq kI$, $y \in G$, and $p \in [1, \infty)$, then

$$\left(\int_G \left[\frac{p_T(xy^{-1})}{p_T(x)} \right]^p p_T(x) dx \right)^{1/p} \leq \exp \left(\frac{c(kT)(p-1)}{2T} d^2(e, y) \right).$$

Proof: Let

- $g \in C^1([0, 1], G)$ such that $g(0) = e \in G$ and $g(1) = y \in G$
- $A_t := L_{g(t)*}^{-1} \dot{g}(t) \in \mathfrak{g}$ and $X_t := \tilde{A}_t \in \Gamma(TG)$.
- So $S_t(x) = xg(t)$.
- $J_1(x) := \frac{d(S_1)_*\mu}{d\mu}(x) = \frac{p_T(xy^{-1})}{p_T(x)}$
- $W_t = -\tilde{A}_t \ln p_T$ is the μ -divergence of $X_t = \tilde{A}_t$.
- $r(0) = 1$, $r(1) = p$ and $\dot{r}(t) > 0$ for $0 < t < 1$.



$$A_t = L_{g(t)*}^{-1} \dot{g}(t) \text{ and } \ell(g) = \int_0^1 |A_t|_g dt.$$

1. By exponential integrability Theorem 3.3,

$$\mathcal{B}(\lambda W_t) = \ln \left(\int_G e^{\lambda W_t} d\mu \right) \leq c\lambda^2 |A_t|_{\mathfrak{g}}^2.$$

2. By extended Galaz-Fontes, Gross, Sontz Theorem 9.2,

$$\left(\int_G \left[\frac{p_T(xy^{-1})}{p_T(x)} \right]^{p'} p_T(x) dx \right)^{1/p'} = \|J_1\|_{p'} \leq e^{\Lambda(r)}, \quad (14)$$

where

$$\begin{aligned} \Lambda(r) &= \int_0^1 \frac{\dot{r}(t)}{r^2(t)} \mathcal{B} \left(\frac{r(t)}{\dot{r}(t)} W_t \right) dt \\ &\leq c \int_0^1 \frac{\dot{r}(t)}{r^2(t)} \frac{r^2(t)}{\dot{r}^2(t)} |A_t|_{\mathfrak{g}}^2 dt = c \int_0^1 \frac{|A_t|_{\mathfrak{g}}^2}{\dot{r}(t)} dt, \end{aligned}$$

3. $\Lambda(r)$ is minimized by taking

$$r(t) = 1 + \kappa \int_0^t |A_\tau|_{\mathfrak{g}} d\tau$$

with

$$\kappa = (p-1) \left(\int_0^1 |A_\tau|_{\mathfrak{g}} d\tau \right)^{-1}.$$

With this choice of r ,

$$\Lambda(r) = \frac{c(kT)(p' - 1)}{2T} \left(\int_0^1 |A_t|_g dt \right)^2 \quad (15)$$

$$= \frac{c(kT)(p' - 1)}{2T} \ell^2(g) \quad (16)$$

4. Combining Eqs. (14) and Eq. (15) and then optimizing over the paths, g , joining e to y gives the result (with $p \rightarrow p'$).

Q.E.D.

A Conjecture

Conjecture 11.1. Let (M, g) be a complete Riemannian manifold such that $\text{Ric} \geq kI$ for some $k \in \mathbb{R}$. Then for every $y, z \in M$ and $p \in [1, \infty)$,

$$\left(\int_G \left[\frac{p_T(y, x)}{p_T(z, x)} \right]^p p_T(z, x) dx \right)^{1/p} \leq \exp \left(\frac{c(kT)(p-1)}{2T} d^2(y, z) \right) \quad (17)$$

where $c = c_T(k)$ is defined as in Eq. (2), $p_t(x, y)$ is the heat kernel on M and $d(y, z)$ is the Riemannian distance from x to y for $x, y \in M$.

On the proof of the Jacobian Estimates

Lemma 12.1 (*Young's Inequality*). Let $W \in L^1(\mu)$ and $f \geq 0$ be a bounded measurable function. Then, for all $s > 0$,

$$\int_M W f d\mu \leq s \int_M f \ln \frac{f}{\mu(f)} d\mu + s\mathcal{B}(W/s) \int_M f d\mu \quad (18)$$

where

$$\mathcal{B}(W) := \ln(\mu(e^W)) = \ln\left(\int_M e^W d\mu\right).$$

Proof: Applying Young's inequality ($xy \leq e^x + y \ln y - y$ for $x \in \mathbb{R}$ and $y \geq 0$) with $x = W$ and $y = f$ and then integrating gives;

$$\int_M W f d\mu \leq \int_M e^W d\mu + \int_M [f \ln f - f] d\mu.$$

Replacing f by λf with $\lambda > 0$ and then optimizing the result in λ implies

$$\int_M W f d\mu \leq \int_M f \ln \frac{f}{\mu(f)} d\mu + \mathcal{B}(W) \int_M f d\mu. \quad (19)$$

Finish the proof by replacing W by W/s .

Q.E.D.

Definition 12.2. The μ -divergence of a smooth vector field, X , on M is the function $W = W_X$ defined by

$$\int_M X\varphi d\mu = \int_M \varphi W d\mu, \text{ for all } \varphi \in C_c^1(M).$$

Proposition 12.3. Let $W_t := W_{X_t}$ be the μ -divergence of X_t , $h \in C^1(M, [0, \infty))$, $h_t := h \circ S_t^{-1}$, and $r \in C^1((0, \tau), (1, \infty))$. Then for any $s > 0$ we have

$$\frac{d}{dt} \ln \|h_t\|_{r(t)} \geq \left(\frac{\dot{r}}{r} - s \right) \int_M \frac{h_t^r}{\|h_t\|_r^r} \left(\ln \frac{h_t}{\|h_t\|_r} \right) d\mu - \frac{s}{r} \mathcal{B}(s^{-1}W_t). \quad (20)$$

Proof: Let $r = r(t)$ and $\dot{r} = \dot{r}(t)$, then by pure calculation;

$$\frac{d}{dt} \ln \|h_t\|_{r(t)} = \frac{\dot{r}}{r} \int_M \frac{h_t^r}{\|h_t\|_r^r} \left(\ln \frac{h_t}{\|h_t\|_r} \right) d\mu - \frac{1}{r} \int_M W_t \left(\overbrace{\frac{h_t^r}{\|h_t\|_r^r}}^f \right) d\mu.$$

Using the estimate in Eq. (18) with $W = W_t$ and $f = \frac{h_t^r}{\|h_t\|_r^r}$ then gives Eq. (20). Q.E.D.

Remark 3. Taking $s = \dot{r}/r$ in Eq. (20) shows;

$$\frac{d}{dt} \ln \|h_t\|_{r(t)} \geq -\frac{\dot{r}}{r^2} \mathcal{B}(s^{-1}W_t). \quad (21)$$

The following theorem is the extension of [Galaz-Fontes *et al.*, 2001, Theorem 2.14] from time independent vector fields to time dependent vector fields.

Theorem 12.4 (*Jacobian Estimate*). Let $p > 1$ and $r \in C([0, \tau], [1, \infty)) \cap C^1((0, \tau), (1, \infty))$ such that $r(0) = 1$, $r(\tau) = p$ and $\dot{r}(t) > 0$ for $0 < t < \tau$, then

$$\|J_\tau\|_{p'} \leq e^{\Lambda(r)}, \quad (22)$$

where $p' := p/(p-1)$ is the conjugate exponent to p and

$$\Lambda(r) = \Lambda_X(r) := \int_0^\tau \frac{\dot{r}(t)}{r^2(t)} \mathcal{B}\left(\frac{r(t)}{\dot{r}(t)} W_t\right) dt. \quad (23)$$

Proof: Integrating Eq. (21);

$$\frac{d}{dt} \ln \|h_t\|_{r(t)} \geq -\frac{\dot{r}}{r^2} \mathcal{B}(s^{-1}W_t),$$

implies

$$\|h \circ S_\tau^{-1}\|_p = \|h_\tau\|_p \geq \|h\|_1 e^{-\Lambda(r)}.$$

Letting $h \rightarrow h \circ S_\tau$ then implies,

$$\int_M h J_\tau d\mu = \|h \circ S_\tau\|_1 \leq \|h\|_p e^{\Lambda(r)}. \quad (24)$$

By the converse to Hölder's inequality, it now follows that Eq. (22) is valid.

Q.E.D.

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A Calculus of Variation Lemma

Lemma 14.1. Let $k \in \mathbb{R}$ and $T > 0$, then

$$\inf \left\{ \int_0^T \dot{\ell}^2(\tau) e^{-k(T-\tau)} d\tau \right\} \leq \frac{k}{e^{kT} - 1} = \frac{c(kT)}{T} \quad (25)$$

where the infimum is taken over all $\ell \in C^1([0, T], \mathbb{R})$ such that $\ell(0) = 0$ and $\ell(T) = 1$.

Proof: By a simple calculus of variation argument, $\ell \in C^1([0, T], \mathbb{R})$ with $\ell(0) = 0$ and $\ell(T) = 1$ is a critical point for the function,

$$K(\ell) := \int_0^T \dot{\ell}^2(\tau) e^{-k(T-\tau)} d\tau, \quad (26)$$

iff $\dot{\ell}(\tau) e^{k\tau}$ is constant in τ . This constraint and the boundary conditions imply that K has a unique critical point at

$$\ell_c(\tau) = \frac{e^{-k\tau} - 1}{e^{-kT} - 1}.$$

Plugging this value of ℓ_c into K then shows $K(\ell_c) = k(1 - e^{-kT})^{-1}$ from which Eq. (25) follows. Q.E.D.