Absolute Continuity of Heat Kernel Measure with Pinned Wiener Measure on Loop groups

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Let \( t > 0 \); \( K \) be a connected compact Lie group equipped with an \( \text{Ad}_K \)-invariant inner product on the Lie Algebra of \( K \). Associated to this data are two measures \( \mu_t^0 \) and \( \nu_t^0 \) on \( \mathcal{L}(K) \) — the space of continuous loops based at \( e \in K \). The measure \( \mu_t^0 \) is pinned Wiener measure with “variance \( t \)” while the measure \( \nu_t^0 \) is a “heat kernel measure” on \( \mathcal{L}(K) \).

The measure \( \mu_t^0 \) is constructed using a \( K \)-valued Brownian motion while the measure \( \nu_t^0 \) is constructed using a \( \mathcal{L}(K) \)-valued Brownian motion.

In this paper we show that \( \nu_t^0 \) is absolutely continuous with respect to \( \mu_t^0 \) and the Radon-Nikodim derivative \( d\nu_t^0/d\mu_t^0 \) is bounded.

1. Introduction

Let \( K \) be a connected compact Lie group, \( \mathfrak{k} \equiv T, K \) be the Lie algebra of \( K \), and \( \langle \cdot, \cdot \rangle = \langle \cdot, \cdot \rangle_{\mathfrak{k}} \) be an \( \text{Ad}_K \)-invariant inner product on \( \mathfrak{k} \). To simplify notation later we will assume that \( K \) is a matrix group. (Since \( K \) is compact, this is no restriction, see for example Theorem 4.1 on p. 136 in [8].)

Example 1.1. As an example, let \( K = \text{SO}(3) \) be the group of \( 3 \times 3 \) real orthogonal matrices with determinant 1. The Lie algebra of \( K \) is \( \mathfrak{k} = \text{so}(3) \), the set of \( 3 \times 3 \) real skew symmetric matrices, and the inner product \( \langle A, B \rangle_{\mathfrak{k}} := -\text{tr}(AB) \) is an example of an \( \text{Ad}_K \)-invariant inner product on \( \mathfrak{k} \).

Elements \( A \in \mathfrak{k} \) will be identified with the unique left invariant vector field on \( K \) agreeing with \( A \) at the identity in \( K \), i.e. if \( f \in C^\infty(K) \) then

\[
Af(x) = \frac{d}{dt}|_0 f(xe^tA).
\]

The path and loop groups on \( K \) are defined by

\[
W(K) \equiv \{ \sigma \in C([0,1] \to K)|\sigma(0) = e\}
\]  

and

\[
\mathcal{L}(K) \equiv \{ \sigma \in W(K)|\sigma(1) = e\}
\]

respectively.

Notation 1.2. The constant path at \( e \) will be denote by \( e \), i.e. \( e(s) = e \) for \( s \in [0,1] \).

\(^1\)Received
\(^2\)This research was partially supported by NSF Grants DMS 96-12651 and DMS 99-71036.
\(^3\)This research was partially supported by NSF Grant DMS 96-12651.

AMS 1991 subject classifications. Primary: 60H07, 58D30 Secondary 58D20

Key words and phrases. Loop groups, heat kernel measures, absolute continuity
Pinned Wiener measure \((\mu_0^e)\) on such a “loop group” (see [21], [24], [4], [17] and Definition 2.11 below) is the law of a \(K\)-valued Brownian motion starting at \(e \in K\) and conditioned to end at \(e \in K\). Heat kernel measure \((\nu_t^0)\) on \(C(K)\) (see [18], [14], and Carson [9, 10] and Definition 2.14 below) is the end point distribution of a \(C(K)\)-valued Brownian motion.” The main theorem (Theorem 2.16) in this paper asserts that \(\nu_t^0\) is absolutely continuous with respect to \(\mu_t^0\) and the Radon-Nikodým derivative \(d\nu_t^0/d\mu_t^0\) is bounded. The proof of this theorem heavily relies on a theorem of Airault and Malliavin (Theorem 2.18 below) which shows that \(\nu_t^0\) solves a heat equation with a potential. A new proof of Theorem 2.18 will be given in Section 6.

One of our motivations for investigating Theorem 2.16 is L. Gross’ logarithmic Sobolev inequality on \((C(K), \mu_t^0)\). To state the inequality, let

\[
\|\text{grad} f\|^2 = \sum_{h \in S_0} (\partial_h f)^2,
\]

where \(S_0\) is an orthonormal basis for \(H_0\) (\(H_0\) is the \(\mathfrak{g}\)-valued Cameron-Martin space in Definitions 3.1) and \(\partial_h\) is a left invariant vector field on \(C(K)\) defined in Definition 3.4. Also let us introduce the following notation. If \(\mu\) is a measure on some measurable space \(\Omega\) and \(f: \Omega \to \mathbb{R}\) is a measurable function, let

\[
\mu(f) = \int_\Omega f d\mu.
\]

L. Gross proves in [20] that there is a constant \(C < \infty\) such that

\[
\int_{C(K)} f^2 \log \frac{f^2}{\mu_t^0(f^2)} d\mu_t^0 \leq C \int_{C(K)} \left\{ \|\text{grad} f\|^2 + V f^2 \right\} d\mu_t^0
\]

where \(V\) is essentially the same potential that appears in the Airault-Malliavin Theorem 2.18 below. It is still an open question as to whether the potential term \(V f^2\) in Eq. (1.4) is necessary or not.

On the other hand, it was shown in Driver and Lohrenz [18] that if \(\mu_t^0\) is replaced by \(\nu_t^0\), the potential term \(V\) is not needed, i.e. there is a constant \(C < \infty\) such that

\[
\int_{C(K)} f^2 \log \frac{f^2}{\nu_t^0(f^2)} d\nu_t^0 \leq C \int_{C(K)} \|\text{grad} f\|^2 d\nu_t^0.
\]

Now Theorem 2.16 below shows that \(Z_t := d\nu_t^0/d\mu_t^0\) is bounded. If one could show that \(Z_t^{-1}\) were also bounded, then the Holley-Stroock lemma (see [22] and Remark 1.20 in [11]) along with Eq. (1.5) would imply that Eq. (1.4) holds without the \(V f^2\) term. It is almost certainly seems too much to expect that \(Z_t\) is bounded from below in general. (It is not even known if \(Z_t > 0\), \(\mu_t^0\)-a.s., when \(K\) is non-abelian.) So the authors do not expect this line of reasoning to work without modification. Nevertheless, better knowledge of the density \(Z_t\) may be useful in determining if potential is needed in Eq. (1.4).
1.1. Conjecture on equivalence. Let us end this introduction with the following conjecture.

Conjecture. If $K$ is simply connected (so that $\mathcal{L}(K)$ has only one connected component) then $Z_t > 0$, $\mu^0_t$ - a.s. That is to say $\mu^0_t$ is absolutely continuous relative to $\nu^0_t$. If $K$ is not simply connected, then we expect that $\mu^0_t$ is absolutely continuous relative to a sum of left translates of $\nu^0_t$ by finite energy loops from each homotopy class.

The explicit calculations in Section 7 shows that the conjecture is true for $K = \mathbb{R}^d$ and $K = S^1$, see Lemma 7.1 and Proposition 7.5. Moreover, the results in Srimurthy [32] also support the conjecture. Let $\mathcal{F}_\alpha$ be the $\sigma$ - algebra consisting of the measurable sets in $W(K)$ depending only on the portion of the paths in $W(K)$ over the interval $[0,\alpha]$, see Definition 2.5 below. Srimurthy proves that $\mu^0_t$ and $\nu^0_t$ are equivalent on $\mathcal{F}_\alpha$ for any $\alpha < 1$. Of course these $\sigma$ - algebras are not able to detect the homotopy classes in $\mathcal{L}(K)$ and it is certainly not true that $\mu^0_t$ is absolutely continuous with respect to $\nu^0_t$ if $K$ is not simply connected. This is because pinned Wiener measure $\mu^0_t$ charges all of the homotopy classes of $K$ while the heat kernel measure $\nu^0_t$ only charges the trivial homotopy class.

2. Notation and Statements of Results

2.1. Brownian Sheets

Definition 2.1 ($\mathbb{R}$ - valued Brownian Sheet). Let $\{\beta(t,s)\}_{0 \leq s \leq t, 0 \leq t < \infty}$ be a $\mathfrak{t}$ - valued Brownian sheet and $\{\chi(t,s)\}_{0 \leq s \leq t, 0 \leq t < \infty}$ be a $\mathfrak{t}$ - valued Brownian bridge sheet defined on some probability space $(\Omega, \mathcal{G}, P)$. To be more precise, let $s \wedge \sigma \equiv \min(s, \sigma)$, $G_0(s, \sigma) = s \wedge \sigma - \sigma s$, $\beta^A(t,s) = \langle A, \beta(t,s) \rangle$ and $\chi^A(t,s) = \langle A, \chi(t,s) \rangle$. Then we are assuming the $\beta$ and $\chi$ are centered Gaussian random fields with covariance functions

\[ \mathbb{E}[\beta^A(t,s)\beta^B(\sigma,\tau)] = \langle A,B \rangle(t \wedge \tau)(s \wedge \sigma) \]

for all $s,\sigma, t, \tau \in [0,\infty)$ and $A, B \in \mathfrak{t}$ and

\[ \mathbb{E}[\chi^A(t,s)\chi^B(\sigma,\tau)] = \langle A,B \rangle(t \wedge \tau)G_0(s,\sigma) \]

for all $s,\sigma \in [0,1]$, $t, \tau \in [0,\infty)$ and $A, B \in \mathfrak{t}$. (Here and in the sequel we will use $\mathbb{E}$ to denote the expectation relative to the measure $P$.)

It is well known that $\beta(t,s)$ and $\chi(t,s)$ may be chosen to have continuous sample paths, see for example the discussion after the proof of Corollary 1.3 in [34]. This fact may also be proved by abstract Wiener space considerations, see Remark 3.3 in [15]. So in the sequel we will assume that $(t,s) \rightarrow \beta(t,s)$ and $(t,s) \rightarrow \chi(t,s)$ are continuous processes.

\[ ^4 \text{Note added in proof. This conjecture is now known to be true, see Aida and Driver [1]. The proof is a combination of the results of this paper, Gross’ ergodicity result in [21] and Malliavin’s [23] quasi-invariance theorem for pinned Wiener measure on loop groups. However, the results in [1] still do not give any lower bound estimates for } Z_t. \]
Definition 2.2. A $\mathbb{t}$-valued process $\{B_s\}$ is said to be a Brownian motion with variance $t$ if $\sqrt{t}B_s$ is a standard $\mathbb{t}$-valued Brownian motion. Alternatively, $B$ may be described using Lévy's characterization (see for example Theorem 39 on p.80 in [23]) of Brownian motion, by requiring $\{B_s\}$ to be a mean zero martingale with quadratic co-variations given by $dB_s^CdB_s^D = t(C,D)ds$ for all $C, D \in \mathbb{t}$.

Remark 2.3. Notice that for fixed $s ; t \mapsto \beta(t,s)$ and $t \mapsto \chi(t,s)$ are $\mathbb{t}$-valued Brownian motions with variance $s$ and $C_0(s,s)$ respectively. This follows by the independent increments of these processes in the $t$ variable, Lemma 8.1 of the Appendix, and Definition 2.2. Similarly for fixed $t ; s \mapsto \beta(s,t)$ is a $\mathbb{t}$-valued Brownian motion with variance $t$. The process $s \mapsto \chi(t,s)$ is a Brownian Bridge for $0 \leq s \leq 1$ with quadratic co-variation given by $\chi^A(t,ds)\chi^B(t,ds) = t(A,B)ds$, see Remark 2.12 below.

Definition 2.4 (Cylinder Functions). For $0 \leq s \leq 1$, let $\pi_s : W(K) \to K$ be the projection map $\pi_s(\sigma) = \sigma(s)$. More generally if

(2.3) \[ \mathbb{P} = \{0 = s_0 < s_1 < s_2 < \ldots < s_n < 1\} \]

is a partition of $[0,1]$, let $s_{n+1} = 1$ by convention and let $\pi_\mathbb{P} : W(K) \to K^n$ be given by

(2.4) \[ \pi_\mathbb{P}(\sigma) = (\sigma(s_1), \sigma(s_2), \ldots, \sigma(s_n)). \]

A cylinder function $f$ on $W(K)$ or $\mathcal{L}(K)$ is a function of the form $f = F \circ \pi_\mathbb{P}$ for some partition $\mathbb{P}$ and some measurable function $F : K^n \to \mathbb{R}$. The function $f$ is said to be bounded (smooth) provided that $F$ is bounded (smooth).

Definition 2.5. For $s \in [0,1]$, let $\mathcal{F}_s$ denote the $\sigma$-algebra on $W(K)$ generated by the smooth cylinder functions of the form $f = F \circ \pi_\mathbb{P}$ where $\mathbb{P}$ runs through partitions as in Eq. (2.3) with $s_n \leq s$. We will write $\mathcal{F}$ for $\mathcal{F}_1$.

The $\sigma$-algebra, $\mathcal{F}$, is the same as the Borel $\sigma$-algebra on $W(K)$, where $W(K)$ is equipped with topology of uniform convergence relative to a metric on $K$ derived from a Riemannian metric on $TK$.

Remark 2.6. For notational simplicity when working on $\mathcal{L}(K)$, we have defined $\pi_\mathbb{P}$ as in Eq. (2.4) rather than by $\pi_\mathbb{P}(\sigma) = (\sigma(s_1), \sigma(s_2), \ldots, \sigma(s_n), \sigma(s_{n+1}))$ which would be more natural on $W(K)$. This results in a slightly smaller class of cylinder functions, but this is of no significance for our purposes.

The next result is well known, but we include it for the reader’s convenience.

Lemma 2.7. Suppose that $Q$ is a finite measure on $(W(K), \mathcal{F})$ and $1 \leq p < \infty$. Then the smooth cylinder functions are dense in $L^p(W(K), \mathcal{F}, Q)$.
Proof of Lemma 2.7. Let \( \mathcal{M} \) denote the smooth cylinder functions and \( \mathcal{H} \) denote those functions in the \( L^p(W(K), \mathcal{F}, Q) \) closure of \( \mathcal{M} \) which are also bounded. Then \( \mathcal{H} \) is a vector space containing the constant functions and which clearly satisfies the property; if \( \{f_n\}_{n=1}^\infty \) is a sequence of functions in \( \mathcal{H} \) such that \( 0 \leq f_1 \leq f_2 \leq f_3 \leq \ldots \), and \( f := \lim_{n \to \infty} f_n \) is bounded, then \( f \in \mathcal{H} \). Since \( \mathcal{M} \) is closed under multiplication, we may apply the monotone class theorem (see Theorem 8 on p. 7 in [28]) to conclude \( \mathcal{H} \) contains all bounded \( \mathcal{F} = \sigma(\mathcal{M}) \) measurable functions. Since (by the dominated convergence theorem) \( \mathcal{H} \) is dense in \( L^p(W(K), \mathcal{F}, Q) \), we are done.

2.2. \( K \)-valued Brownian motion and Wiener measures

Definition 2.8 (Wiener Measure on \( W(K) \)). Fix \( t > 0 \), let \( \{g_s\}_{s \in [0,1]} \) denote the solution to the stochastic differential equation
\[
dg_s = g_s \beta(t, \delta s) \text{ with } g_0 = e \in K,\]
where \( \beta(t, \delta s) \) denotes the Stratonovich differential of the Brownian motion \( s \to \beta(t, s) \). The Wiener measure with variance \( t \) on \( \mathcal{F} \) is \( \mu_t := \text{Law}(g_s) \).

Let \( \mathfrak{e}_0 \subset \mathfrak{e} \) be an orthonormal basis for \( \mathfrak{e} \), and \( \Delta_K \) be the second order elliptic operator,
\[
\Delta_K = \sum_{A \in \mathfrak{e}_0} A^2.
\]
Since \( K \) is compact and hence uni-modal, \( \Delta_K \) is the Laplace Beltrami operator for the left invariant Riemannian metric on \( K \) determined by \( \langle \cdot, \cdot \rangle \) on \( \mathfrak{e} = T_K K \), see for example Remark 2.2 in [16]. Using It\'s lemma, one easily shows that \( \{g_s\}_{s \in [0,1]} \) is a diffusion process on \( K \) with generator \( \frac{1}{2} \Delta_K \). Such a \( K \)-valued process will be called a Brownian motion on \( K \) with variance \( t \).

Definition 2.9 (Heat Kernel on \( K \)). Let \( p^K_t \) denote the smooth function of \( K \) such that \( \text{Law}(g_1) = p^K_t(x)dx \), where \( dx \) denotes normalized Haar measure on \( K \).

The function \( p^K_t \) is the convolution kernel for the heat operator \( e^{\Delta_K/2} \). In particular, \( (t, x) \to p^K_t(x) \) is a smooth positive function such that for any \( f \in C(K) \), the function \( u \) defined by
\[
u(t, x) \equiv \int_K f(y) p^K_t(x^{-1}y) \, dy \text{ for } (t, x) \in (0, \infty) \times K
\]
satisfies the heat equation
\[
\partial_t u = \frac{1}{2} \Delta_K u \text{ with } \lim_{t \to 0} u(t, x) = f(x)
\]
where \( \partial_t = \partial/\partial t \).
Remark 2.10. It is well known that \( p^K_t(x) = p^K_t(x^{-1}) \) for all \( x \in K \), see for example Item 2 of Proposition 3.1 in [16]. It also well known that \( p^K_t \) is a class function, i.e.

\[
(2.7) \quad p^K_t(xy) = p^K_t(yx) \quad \text{for all} \quad x, y \in K.
\]

This is a consequence of the fact that \( \Delta_K \) is a bi-invariant differential operator because of the \( \text{Ad}_K \) – invariance of \( \langle \cdot, \cdot \rangle \). Thus for all bounded measurable functions \( f \) on \( K \),

\[
\int_K f(y)p^K_t(x^{-1}y) \, dy = \int_K f(xy)p^K_t(y) \, dy
\]

\[
= \left( e^{t\Delta_K/2} f \circ L_x \right)(e) \quad \left( e^{t\Delta_K/2} f \right)(x) = \left( e^{t\Delta_K/2} f \circ R_x \right)(e)
\]

\[
= \int_K f(yx)p^K_t(y) \, dy = \int_K f(y)p^K_t(y^{-1}) \, dy,
\]

where \( L_x \) and \( R_x \) denote left and right multiplication by \( x \in K \) respectively. The last displayed equation implies Eq. (2.7).

By the Markov property of \( g \), and the previous comments, if \( f \) is a bounded cylinder function of the form \( f(\sigma) = F \circ \pi_x \) where \( \mathbb{P} \) is as in Eq. (2.3), then

\[
(2.8) \quad \mu_t(f) \equiv \int_{K^n} F(x_1, \cdots, x_n) \prod_{i=1}^n p^K_{\Delta_i s}(x_{i-1}^{-1}x_i) \, dx_1dx_2\cdots dx_n,
\]

where \( x_0 := e \) and \( \Delta_is = s_i - s_{i-1} \).

Definition 2.11 (Doob’s Construction of Pinned Wiener Measure).

Pinned Wiener measure, \( \mu^0_t \), on \( W(K) \) with variance \( t \), is the unique measure on \( \mathcal{F} \) such that if \( f \) is a bounded \( \mathcal{F}_\alpha \) measurable function for some \( \alpha \in (0, 1) \), then

\[
\mu^0_t(f) \equiv \frac{1}{p^K_t(e)} \mu_t(f_{p^K_t(\alpha)}(\pi_\alpha)).
\]

In particular if \( f \) is a bounded cylinder function \( f \) of the form \( f(\sigma) = F \circ \pi_x \) where \( \mathbb{P} \) is as in Eq. (2.3), then

\[
(2.9) \quad \mu^0_t(f) \equiv \int_{K^n} F(x)\rho^\mathbb{P}(t, x)dx,
\]

where \( x = (x_1, \cdots, x_n) \), \( dx = dx_1dx_2\cdots dx_n \) is normalized Haar measure on \( K^n \) and

\[
(2.10) \quad \rho^\mathbb{P}(t, x) := \frac{1}{p^K_t(e)} \prod_{i=1}^{n+1} p^K_{t(s_i - s_{i-1})}(x_{i-1}^{-1}x_i)
\]

where by convention \( x_0 = x_{n+1} = e \).
The existence of the probability measure $\mu_t^0$ and the fact that $\mu_t^0(\mathcal{L}(K)) = 1$ is well known. A proof may be found, for example, in Theorem 2.3 in [12]. To apply this theorem, the reader should take the covariant derivative $\nabla$ appearing in Theorem 2.3 in [12] to be the unique one for which left invariant vector fields on $K$ are covariantly constant.

Remark 2.12. In Remark 2.3 it was asserted that the process $s \rightarrow \chi(t, s)$ is a Brownian bridge with quadratic co-variation given by $\chi^A(t, ds)\chi^C(t, ds) = t(A, C)ds$, that is to say $\text{Law}(\chi(t, \cdot))$ is pinned Wiener measure on $\mathcal{L}(\mathfrak{t})$ with variance $t$. To check this let $p_t(x) = (2\pi t)^{-\dim k/2}e^{-1/(2t)}$ be the Euclidean heat kernel on $\mathfrak{t}$. Then for a cylinder function $f$ on $\mathcal{L}(\mathfrak{t})$ based on a partition $\mathbb{P} = \{0 = s_0 < s_1 < s_2 < \ldots < s_n < 1\}$, we must show that

$$\mathbb{E}f(\chi(t, \cdot)) = \mathbb{E} \left[ f(B) \frac{p_t(1-\sigma)(B_\sigma)}{p_t(0)} \right],$$

where $B_s = \beta(t, s) - a \mathfrak{t}$-valued Brownian motion with variance $t$.

Proof of Remark 2.12. To prove Eq. (2.11), let $Z_s = p_t(0)^{-1}p_t(1-s)(\beta(t, s))$ for $0 \leq s < 1$, then by Itô’s lemma and the fact that

$$\frac{\partial}{\partial s}p_t(1-s)(x) = -\frac{1}{2}\Delta p_t(1-s)(x) \quad \text{and} \quad \nabla \log p_t(1-s)(x) = -\frac{1}{t(1-s)}x$$

we have

$$dZ_s = -Z_s \frac{1}{t(1-s)}\langle B_s, dB_s \rangle \text{ with } Z_0 = 1.$$

By Girsanov’s theorem (see for example Theorem 20 on p.109 in [28])

$$M_s := B_s - \int_0^s \frac{1}{Z_r}dZ_r dB_r = B_s + \int_0^s \frac{1}{1-r}B_r dr$$

is a martingale on $[0, \alpha]$ relative to the measure $Z_\alpha P$. Since $M$ has the same quadratic variation as $\beta(t, \cdot)$, by Lévy’s criteria, $M$ is a $\mathfrak{t}$-valued Brownian motion with variance $t$ under the measure $Z_\alpha P$. Interpreting (2.12) as stochastic differential equation for $B$,

$$dB = dM - \frac{1}{1-s}B_s ds \text{ with } B_0 = 0,$$

we find by variation of parameters that

$$B_s = \int_0^s e^{-\int_r^s \frac{1}{1-r'}dr'}dM_r = \int_0^s \int_0^s \frac{1-s}{1-r}dM_r.$$

This shows that, under $Z_\alpha P$, $\{B_s\}_{0 \leq s \leq \alpha}$ is still a Gaussian process. Moreover, for $0 \leq \sigma \leq s \leq \alpha$,

$$\mathbb{E}[B_s^C B_r^D Z_\alpha] = \mathbb{E}\left[ \left( \int_0^s \frac{1-s}{1-r}dM_r \right) \left( \int_0^\sigma \frac{1-\sigma}{1-r}dM_r^D \right) Z_\alpha \right]$$

$$= t(1-s)(1-\sigma)(C, D) \int_0^{\sigma} \frac{1}{(1-r)^2} dr.$$
which is the same covariance function as $\chi(t, \cdot)$. Therefore $\{B_s\}_{0 \leq s \leq \alpha}$ under the measure $Z_\alpha P$ has the same law as $\{\chi(t, s)\}_{0 \leq s \leq \alpha}$ under the measure $P$. This is the assertion in Eq. (2.11).

2.3. Heat kernel measure on $W(K)$ and $L(K)$ In this section we are going to define heat kernel measures on $W(K)$ and $L(K)$ by formally replacing $K$ from the previous section by $W(K)$ and $L(K)$ respectively. Following Malliavin [25], we have the following theorem.

**Theorem 2.13 (Brownian Motion on $W(K)$ and $L(K)$).** There are jointly continuous solutions $(\Sigma(t, s))$ and $(\Sigma^0(t, s))$ to the stochastic differential equations:

\begin{align}
(2.13) \quad \Sigma(\delta t, s) &= \Sigma(t, s) \beta(\delta t, s) \text{ with } \Sigma(0, s) = e \quad \forall s \in [0, \infty), \\
(2.14) \quad \Sigma^0(\delta t, s) &= \Sigma^0(t, s) \chi(\delta t, s) \text{ with } \Sigma(0, s) = e \quad \forall s \in [0, 1].
\end{align}

As before $\beta(\delta t, s)$ denotes the Stratonovich differentials of the processes $t \to \beta(t, s)$ and similarly for $\Sigma(\delta t, s)$, $\Sigma^0(\delta t, s)$, and $\chi(\delta t, s)$.

**Proof of Theorem 2.13.** Such results may be found in Baxendale, [5], Malliavin [25], or in Theorem 3.8 in Driver [14]. The last two references cover the $L(K)$ case, however the proof of the $W(K)$ case is the same, just replace $G_0(s, \alpha)$ by $s \land \alpha$ throughout.

**Definition 2.14 (Heat Kernel Measures on $W(K)$ and $L(K)$).** The measures $\nu_t = \text{Law}(\Sigma(t, \cdot))$ and $\nu^0_t = \text{Law}(\Sigma^0(t, \cdot))$ are called heat kernel measures on $W(K)$ and $L(K)$ respectively. So $\nu_t$ and $\nu^0_t$ are determined by

\begin{align}
(2.15) \quad \nu_t(f) &= \mathbb{E} f(\Sigma(t, \cdot)) \text{ and } \nu^0_t(f) = \mathbb{E} f(\Sigma^0(t, \cdot))
\end{align}

for all bounded $\mathcal{F}$ measurable $f$. Notice that $\nu_0^0(L(K)) = 1$ because $\Sigma^0(t, 0) = \Sigma^0(t, 1) = e$, $P$-almost surely.

Corollary 3.10 below justifies calling $\nu_t$ and $\nu^0_t$ heat kernel measures.

2.4. Statement of Results The following theorem is Lemma 1 in Airault and Malliavin [2].

**Theorem 2.15.** Let $t > 0$, then $\nu_t = \mu_t$ on $W(K)$, i.e. heat kernel measure at time $t$ and Wiener measure with variance $t$ are the same on $W(K)$. 
This theorem is also proved in Lemma 3.3 of Srinurthy [32]. Since this theorem is crucial to the rest of the paper, we will give a proof in Section 4 below. The following theorem is the main result of the paper.

**Theorem 2.16.** Let $t > 0$, then $\nu_t^0 \ll \mu_t^0$, i.e. heat kernel measure at time $t$ is absolutely continuous relative to pinned Wiener measure with variance $t$. Moreover, the Radon-Nikodym derivative, $d\nu_t^0/d\mu_t^0$, satisfies the bound

$$\frac{d\nu_t^0}{d\mu_t^0} \leq e^{C_t},$$

where

$$C_t \equiv \log \left( (2\pi t)^{\frac{1}{2}} \dim \mathfrak{p} \mu_t^0(e) \right).$$

(Standard heat kernel asymptotics shows that $\lim_{t \to 0} C_t = 0$, see Lemma 6.1 below.)

The proof of this theorem (given in Section 6) will be a combination of the maximum principle along with a theorem of Airault and Malliavin [3]. In order to state the Airault-Malliavin theorem, let us recall that the coordinate process $\pi_s : \mathcal{L}(K) \to K$ (see Definition 2.4) is a semi-martingale relative to pinned Wiener measure, $\mu_t^0$, see for example Bismut [7] or Theorem 2.3 in [13]. Hence we may define the $\mathfrak{p}$-valued semi-martingale $\{b_s\}_{0 \leq s \leq 1}$ by

$$b_s := \int_0^s \pi_r^{-1} \delta \pi_r.$$

**(Remarks 2.17.** i) Technically speaking the stochastic integral in Eq. (2.17) depends on the measure $\mu_t^0$ and in particular on $t > 0$. So a more appropriate notation would be to display this $t$ dependence and write $b^t_s$ for the $\mu_t^0$ - a.e. defined stochastic integral $\int_0^s \pi_r^{-1} \delta \pi_r$. Since we will only need the process $b_s$ for one fixed value of $t$, we will stick with the notation in Eq. (2.17).

ii) Gross shows (see Lemma 4.8 and Remark 4.9 in [20]) that $b_1 \in L^p(\mathcal{L}(K), \mu_1^0)$ and that $b_s \to b_1$ in $L^p(\mathcal{L}(K), \mu_1^0)$ as $s \to 1$ for all $1 \leq p < \infty$.

**Theorem 2.18 (Airault & Malliavin).** Let $V_t : \mathcal{L}(K) \to \mathbb{R}$ be the “potential,”

$$V_t = \frac{1}{2t} \left| b_1 \right|^2_t - c_t$$

where $b_1$ is defined in Eq. (2.17) and

$$c_t \equiv \frac{dC_t}{dt} = \frac{\dim \mathfrak{p}}{2t} + \partial_t \log \mu_t^0(e).$$

Then for any smooth cylindrical function $f : \mathcal{L}(K) \to \mathbb{R}$ (see Definition 2.4)

$$\partial_t \mu_0^0(f) = \mu_0^0 \left[ \left( \frac{1}{2} \Delta_{\mathcal{L}(K)} + V_t \right) f \right],$$

where $\Delta_{\mathcal{L}(K)}$ is the generator of the process $\Sigma^0(t, \cdot)$, see Definition 3.6 and Proposition 3.9 below.
We will give a simplified (in our view) proof of this theorem in Section 5. The proof relies on Theorem 2.15 and integration by parts on \((W(K), \mu_t)\).

3. Generators of \(\Sigma(t, \cdot)\) and \(\Sigma^0(t, \cdot)\) Much of the material in this section may be found in [18] and [14]. Nevertheless, in order to introduce the notation and for the readers convenience we will summarize some of the results in these papers.

3.1. Cameron-Martin spaces

**Definition 3.1.** Given a continuous function \(h : [0, 1] \to \mathfrak{k}\) define

\[
(h, h)_H = \begin{cases} 
\int_0^1 |h'(s)|^2 \, ds & \text{if } h \text{ is absolutely continuous} \\
\infty & \text{otherwise}.
\end{cases}
\]

The Cameron–Martin space of \(\mathfrak{k}\) is

\[H \equiv \{h \in C([0, 1] \to \mathfrak{k}) | h(0) = 0 \text{ and } (h, h) < \infty\}\]

which we equip with the inner product

\[(h, k) = \int_0^1 \langle h'(s), k'(s) \rangle ds.
\]

The pinned Cameron–Martin space is

\[H_0 \equiv \{h \in H(\mathfrak{k}) | h(1) = 0\}
\]

which is a closed subspace of \(H\). (The Hilbert spaces \(H\) and \(H_0\) are to be thought of as the “Lie algebras” to the groups \(W(K)\) and \(L(K)\).)

**Notation 3.2.** Let \(S \subset H\) and \(S_0 \subset H_0\) be orthonormal bases for \(H\) and \(H_0\) respectively.

**Lemma 3.3.** Let \(\mathfrak{k}_0 \subset \mathfrak{k}\) be an orthonormal basis for \(\mathfrak{k}\), \(G(s, t) = s \wedge t\) and \(G_0(s, t) \equiv s \wedge t - st\) for all \(s, t \in [0, 1]\). Then

\[
(3.1) \quad \sum_{h \in S} h(s) \otimes h(t) = G(s, t) \sum_{A \in \mathfrak{k}_0} A \otimes A \in \mathfrak{k} \otimes \mathfrak{k}.
\]

\[
(3.2) \quad \sum_{h \in S_0} h(s) \otimes h(t) = G_0(s, t) \sum_{A \in \mathfrak{k}_0} A \otimes A \in \mathfrak{k} \otimes \mathfrak{k}.
\]

**Proof of Lemma 3.3.** Let \(A, B \in \mathfrak{k}\). Since \(G(t, \cdot)B\) and \(G(s, \cdot)A\) are in \(H\),

\[
(3.3) \quad (G(t, \cdot)B, G(s, \cdot)A) = \sum_{h \in S} (G(t, \cdot)B, h)(h, G(s, \cdot)A)
\]

where the sum is absolutely convergent. By the fundamental theorem of calculus, \(G\) satisfies the reproducing property,

\[
\int_0^1 \partial_s G(t, s) h'(s) ds = h(t) \text{ for all } h \in H.
\]
Combined this equation with Eq. (3.3) shows that
\[ G(s,t) = \sum_{h \in S} \langle B, h(t) \rangle \langle h(s), A \rangle \]
which implies Eq. (3.1) since \( A \) and \( B \) are arbitrary. Equation (3.2) is proved similarly, see Lemma 3.8 in [18] for more details.

3.2. Derivatives and Laplacians on \( \mathcal{L}(K) \) and \( W(K) \)

**Definition 3.4 (Left invariant derivatives).** Given \( h \in H \) (or \( H_0 \)) and \( f : W(K) \to \mathbb{R} \) (or \( f : \mathcal{L}(K) \to \mathbb{R} \)) a smooth cylinder function, define
\[
(\partial_h f)(\sigma) := \frac{d}{dt} f(\sigma e^{th}) \text{ for all } \sigma \in W(K) \text{ (} \sigma \in \mathcal{L}(K) \text{)}
\]
where \( \sigma e^{th} \in W(K) \) (\( \sigma e^{th} \in \mathcal{L}(K) \)) is defined by \( (\sigma e^{th}) (s) := \sigma(s)e^{th(s)} \) for \( s \in [0,1] \).

**Remark 3.5.** Suppose that \( f = F \circ \pi_\mathbb{P} \) where \( \mathbb{P} = \{0 = s_0 < s_1 < s_2 < \ldots < s_n < 1\} \) is a partition of \([0,1]\) and \( F : K^n \to \mathbb{R} \) is a smooth function. For \( A \in \mathfrak{t} \) and \( i \in \{1,2,\ldots,n\} \), let
\[
A^{(i)}F(x_1,x_2,\ldots,x_n) = \frac{d}{dt} |_{0} F(x_1,x_2,\ldots,x_{i-1},x_i e^{th},x_{i+1},\ldots,x_n),
\]
so that \( A^{(i)} \) is the action of \( A \) on the \( i^{th} \) variable of \( F \). Then for \( h \in H \) (or \( h \in H_0 \)),
\[
(3.4) \quad \partial_h f = \sum_{i=1}^{n} (h(s_i)^{(i)}F) \circ \pi_\mathbb{P}.
\]
In particular \( \partial_h f \) is still a smooth cylinder function. Therefore the operator \( \partial_h^2 f \) is well defined and is given by
\[
(3.5) \quad \partial_h^2 f = \sum_{i,j=1}^{n} (h(s_j)^{(j)} h(s_i)^{(i)}F) \circ \pi_\mathbb{P}.
\]

**Definition 3.6.** Again suppose that \( f = F \circ \pi_\mathbb{P} \) is a smooth cylinder function as in Definition 2.4. Define the Laplacians on \( W(K) \) and \( \mathcal{L}(K) \) by
\[
\Delta_{W(K)} f \equiv \sum_{h \in S} \partial_h^2 f \text{ and } \Delta_{\mathcal{L}(K)} f \equiv \sum_{h \in S_0} \partial_h^2 f
\]
respectively.
Remark 3.7. Combining Eqs. (3.1), (3.2) and (3.5) we find
\[
\Delta_{W(K)} f = \sum_{h \in S} \sum_{i,j=1}^{n} \left( h(s_j)^{(j)} h(s_i)^{(i)} F \right) \circ \pi_{\mathbb{P}}
\]
(3.6)
\[
= \sum_{A \in \mathcal{E}_0} \sum_{i,j=1}^{n} G(s_i, s_j) \left( A^{(j)} A^{(i)} F \right) \circ \pi_{\mathbb{P}}
\]
and
\[
\Delta_{L(K)} f = \sum_{h \in S_0} \sum_{i,j=1}^{n} \left( h(s_j)^{(j)} h(s_i)^{(i)} F \right) \circ \pi_{\mathbb{P}}
\]
(3.7)
\[
= \sum_{A \in \mathcal{E}_0} \sum_{i,j=1}^{n} G_0(s_i, s_j) \left( A^{(j)} A^{(i)} F \right) \circ \pi_{\mathbb{P}}.
\]

Notation 3.8. Given, \( \mathbb{P} = \{0 = s_0 < s_1 < s_2 < \ldots < s_n < 1\} \), a partition of \([0, 1]\) and \( F \in C^\infty(K^n) \), let
\[
L_{\mathbb{P}} F = \sum_{A \in \mathcal{E}_0} \sum_{i,j=1}^{n} G(s_i, s_j) A^{(j)} A^{(i)} F
\]
(3.8)
and
\[
L_{\mathbb{P}}^0 F = \sum_{A \in \mathcal{E}_0} \sum_{i,j=1}^{n} G_0(s_i, s_j) A^{(j)} A^{(i)} F.
\]
(3.9)

With this notation we may write Eqs. (3.6) and (3.7) as
\[
\Delta_{W(K)} (F \circ \pi_{\mathbb{P}}) = (L_{\mathbb{P}} F) \circ \pi_{\mathbb{P}} \quad \text{and} \quad \Delta_{L(K)} (F \circ \pi_{\mathbb{P}}) = (L_{\mathbb{P}}^0 F) \circ \pi_{\mathbb{P}}.
\]
(3.10)

3.3. Heat equations

Proposition 3.9. The processes \( \Sigma(t, \cdot) \) and \( \Sigma^0(t, \cdot) \) are diffusion processes with \( \Delta_{W(K)} \) and \( \Delta_{L(K)} \) as generators. More precisely, if \( f = F \circ \pi_{\mathbb{P}} \) is a cylinder function as above, then
\[
M_t^f = f(\Sigma(t, \cdot)) - f(\mathbf{e}) - \frac{1}{2} \int_0^t (\Delta_{W(K)} f) (\Sigma(\tau, \cdot)) d\tau
\]
(3.11)
and
\[
N_t^f = f(\Sigma^0(t, \cdot)) - f(\mathbf{e}) - \frac{1}{2} \int_0^t (\Delta_{L(K)} f) (\Sigma^0(\tau, \cdot)) d\tau
\]
(3.12)
are martingales.
Proof of Proposition 3.9. We will only prove Eq. (3.11) since the proof of Eq. (3.12) is completely analogous. Let \( P(t) := P((t, s_1), \ldots, (t, s_n)) \), then \( f(S(t, \cdot)) = F(S_p(t)) \) and by Itô's Lemma we have that
\[
df((t, \cdot)) = dF(S_p(t))
\]
\[
= \sum_{i=1}^{n} \sum_{A \in \mathcal{A}_0} A(i) F(S_p(t))\beta^A(\delta t, s_i)
\]
\[
= \sum_{i=1}^{n} \sum_{A \in \mathcal{A}_0} A(i) F(S_p(t))\beta^A(dt, s_i)
\]
\[
+ \frac{1}{2} \sum_{i, j=1}^{n} \sum_{A, B \in \mathcal{A}_0} B(i) A(j) F(S_p(t))\beta^A(dt, s_i)\beta^B(dt, s_j)
\]
\[
= \sum_{i=1}^{n} \sum_{A \in \mathcal{A}_0} A(i) F(S_p(t))\beta^A(dt, s_i)
\]
\[
+ \frac{1}{2} \sum_{i, j=1}^{n} \sum_{A \in \mathcal{A}_0} G(s_i, s_j) A(i) A(j) F(S_p(t))dt
\]
\[
= \sum_{i=1}^{n} \sum_{A \in \mathcal{A}_0} A(i) F(S_p(t))\beta^A(dt, s_i) + \frac{1}{2} (\Delta W(K)f) (S(t, \cdot))dt.
\]

This shows that \( M^f_t \) is the martingale;
\[
M^f_t = \sum_{i=1}^{n} \sum_{A \in \mathcal{A}_0} \int_0^t A(i) F(S_p(\tau))\beta^A(\tau, s_i) d\tau.
\]

Corollary 3.10. The measures \( \nu_t \) and \( \nu^0_t \) satisfy the heat equations on \( W(K) \) and \( L(K) \) in the following weak sense. If \( f : W(K) \to \mathbb{R} \) is a smooth cylinder function then
\[
\partial_t \nu_t(f) = \frac{1}{2} \nu_t(\Delta W(K)f)
\]
(3.13)

and
\[
\partial_t \nu^0_t(f) = \frac{1}{2} \nu^0_t(\Delta L(K)f).
\]

Proof of Corollary 3.10. Taking expectations of Eq. (3.11) shows that
\[
0 = \mathbb{E}M^f_t = \mathbb{E}f(S(t, \cdot)) - f(e) - \frac{1}{2} \int_0^t \mathbb{E}\left(\Delta W(K)f \right) (S(\tau, \cdot))d\tau
\]
\[
= \nu_t(f) - f(e) - \frac{1}{2} \int_0^t \nu_\tau(\Delta W(K)f) d\tau.
\]

Differentiating this equation in \( t \) proves Eq. (3.13). Eq. (3.14) is proved analogously.
Corollary 3.11 (Heat solution). Suppose that \( u : \mathcal{L}(K) \to \mathbb{R} \) is a smooth cylinder function and let

\[
H(t, \sigma) = \int_{\mathcal{L}(K)} u(\sigma \gamma^{-1}) d\nu_t^0(\gamma),
\]

then

\[
\partial_t H(t, \sigma) = \frac{1}{2} \Delta_{\mathcal{L}(K)} H(t, \sigma) \quad \text{and} \quad \lim_{t \to 0} H(t, \sigma) = u(\sigma)
\]

Proof of Corollary 3.11. For \( \sigma \in \mathcal{L}(K) \), let \( u_\sigma : \mathcal{L}(K) \to \mathbb{R} \) be the cylinder function defined by \( u_\sigma(\gamma) = u(\sigma \gamma^{-1}) \). Notice that for \( h \in H \),

\[
(\partial_h u_\sigma)(\gamma) = \frac{d}{d\epsilon} u_\sigma(\gamma e^{\epsilon h}) = \frac{d}{d\epsilon} u(\sigma e^{-\epsilon h} \gamma^{-1}) = -\partial_h (\sigma \to u_\sigma(\gamma))
\]

and therefore

\[
(\Delta_{\mathcal{L}(K)} u_\sigma)(\gamma) = \Delta_{\mathcal{L}(K)} (\sigma \to u_\sigma(\gamma)).
\]

Thus by Corollary 3.10,

\[
\partial_t H(t, \sigma) = \frac{1}{2} \int_{\mathcal{L}(K)} (\Delta_{\mathcal{L}(K)} u_\sigma)(\gamma) d\nu_t^0(\gamma)
\]

\[
= \frac{1}{2} \int_{\mathcal{L}(K)} \Delta_{\mathcal{L}(K)} (\sigma \to u_\sigma(\gamma)) d\nu_t^0(\gamma)
\]

\[
= \frac{1}{2} \Delta_{\mathcal{L}(K)} \left( \sigma \to \int_{\mathcal{L}(K)} u_\sigma(\gamma) d\nu_t^0(\gamma) \right)
\]

\[
= \frac{1}{2} \Delta_{\mathcal{L}(K)} H(t, \sigma).
\]

Working with the explicitly representation of \( u \) as a cylinder function and using Eq. (3.10), it is easy to justify the interchange of \( \Delta_{\mathcal{L}(K)} \) with the integral in the third equality. This proves the first assertion in Eq. (3.16). The second follows from the dominated convergence theorem and the identity,

\[
H(t, \sigma) = \mathbb{E} \left[ u(\sigma \Sigma^0(t, \cdot)^{-1}) \right],
\]

where \( \Sigma^0(t, s) \) is the process defined in Eq. (2.14) of Theorem 2.13.

4. The path group case In the next subsection we will give a proof of Theorem 2.15. However, before doing this let us record the following trivial Corollary of Theorem 2.15 and Corollary 3.10 above. This corollary will be key to our proof of the Airault Malliavin theorem in Section 5.

Corollary 4.1. The Wiener measure \( \mu_t \) with variance \( t \) satisfies (weakly) the heat equation on \( W(K) \), i.e. if \( f : W(K) \to \mathbb{R} \) is a smooth cylinder function then

\[
\partial_t \mu_t(f) = \frac{1}{2} \mu_t (\Delta_{W(K)} f).
\]
4.1. Proof of Theorem 2.15

Proof of Theorem 2.15. As mentioned in Section 2, the reader may find this theorem in Lemma 1 of Airault and Malliavin [2] or Lemma 3.3 of Srinurthy [32]. It would also be possible to give a proof using two parameter stochastic calculus as developed in Norris [27]. Rather than introduce this machinery, we will give a more pedestrian but perhaps less illuminating proof. Our proof is similar to that in [32].

Let \( \Sigma \) denote the process defined in Theorem 2.13 and \( \mathbb{P} = \{0 = s_0 < s_1 < s_2 < \ldots < s_n < 1\} \) be a partition of \([0, 1]\). Let

\[
U_i(t) := \Sigma(t, s_i) \Sigma(t, s_{i-1})^{-1}
\]

and

\[
B_i(t) := \int_0^t Ad_{\Sigma(t, s_{i-1})} (\beta(\delta \tau, s_i) - \beta(\delta \tau, s_{i-1}))
\]

for \( i = 1, 2, \ldots, n \). By Eq. (2.13) and Itô’s Lemma,

\[
\delta_i \Sigma(t, s)^{-1} = -\beta(\delta t, s) \Sigma(t, s)^{-1}
\]

and therefore

\[
\delta U_i(t) = \Sigma(t, s_i) (\beta(\delta t, s_i) - \beta(\delta t, s_{i-1})) \Sigma(t, s_{i-1})^{-1}
= U_i(t) Ad_{\Sigma(t, s_{i-1})} (\beta(\delta t, s_i) - \beta(\delta t, s_{i-1}))
= U_i(t) \delta B_i(t).
\]

Because, \( t \to \beta(t, s_{i-1}) \) and \( t \to \beta(t, s_i) - \beta(t, s_{i-1}) \) are independent Brownian motions on \( \mathfrak{f} \),

\[
Ad_{\Sigma(t, s_{i-1})} (\beta(\delta t, s_i) - \beta(\delta t, s_{i-1}))
= Ad_{\Sigma(t, s_{i-1})} (\beta(dt, s_i) - \beta(dt, s_{i-1}))
+ \frac{1}{2} Ad_{\Sigma(t, s_{i-1})} [\beta(dt, s_{i-1}), (\beta(dt, s_i) - \beta(dt, s_{i-1}))] \mathfrak{f}
= Ad_{\Sigma(t, s_{i-1})} (\beta(dt, s_i) - \beta(dt, s_{i-1})).
\]

Therefore the Stratonovich differentials in Eq. (4.3) may be replaced by Itô differentials to learn that \( B_i(t) \) is the martingale

\[
B_i(t) := \int_0^t Ad_{\Sigma(t, s_{i-1})} (\beta(d\tau, s_i) - \beta(d\tau, s_{i-1})).
\]

Claim. The processes \( B_1, B_2, \ldots, B_n \) are independent \( \mathfrak{f} \) – valued Brownian motions with variances \( \Delta_i := s_i - s_{i-1} \) for \( i = 1, 2, \ldots, n \).

To prove this claim, let \( C, D \in \mathfrak{f} \), and let \( B^C_j(t) = \langle B_j(t), C \rangle \), \( B^D_j(t) = \langle B_j(t), D \rangle \) and \( \Delta_i \beta(t) := \beta(t, s_i) - \beta(t, s_{i-1}) \). Then because \( \langle \cdot, \cdot \rangle \) is \( Ad_{\Sigma} \) – invariant,

\[
dB^C_i(t) = \langle Ad_{\Sigma(t, s_{i-1})} d\Delta_i \beta(t), C \rangle = \langle d\Delta_i \beta(t), Ad_{\Sigma(t, s_{i-1})}^{-1} C \rangle.
\]
Thus the differential of the quadratic co-variation of $B^C_i$ and $B^D_j$ is given by,
\[
dB^C_i(t)dB^D_j(t) = \langle d\Delta_i \beta(t), Ad_{\Sigma(t,s_{i-1})}^{-1} C \rangle \langle d\Delta_j \beta(t), Ad_{\Sigma(t,s_{j-1})}^{-1} D \rangle
\]
\[
= \sum_{A \in \mathcal{E}_0} \langle A, Ad_{\Sigma(t,s_{i-1})}^{-1} C \rangle \langle A, Ad_{\Sigma(t,s_{j-1})}^{-1} D \rangle d\Delta_i \beta^A(t) d\Delta_j \beta^A(t)
\]
\[
= \delta_{ij} \sum_{A \in \mathcal{E}_0} \langle A, Ad_{\Sigma(t,s_{i-1})}^{-1} C \rangle \langle A, Ad_{\Sigma(t,s_{j-1})}^{-1} D \rangle \Delta_i s dt
\]
\[
= \delta_{ij} \langle Ad_{\Sigma(t,s_{i-1})}^{-1} C, Ad_{\Sigma(t,s_{j-1})}^{-1} D \rangle \Delta_i s dt
\]
(4.5)
wherein the third equality we have used: i) $\Delta_i \beta^A(\cdot) = \beta^A(\cdot, s_i) - \beta^A(\cdot, s_{i-1})$ and $\Delta_j \beta^A(\cdot) = \beta^A(\cdot, s_j) - \beta^A(\cdot, s_{j-1})$ are independent if $i \neq j$ and $\Delta_i \beta^A(\cdot)$ is a $t$-valued Brownian motion with variance $\Delta_i s$. In the last equality we again have used the $Ad_K$ invariance of $\langle \cdot, \cdot \rangle$. Eq. (4.5) along with Lévy’s criteria proves the claim.

Since the $U_i$’s in Eq. (4.2) satisfy Eq. (4.4), the claim implies that $U_1(t), U_2(t), \ldots, U_n(t)$ are independent $K$-valued Brownian motion with variance $\Delta_1 s, \Delta_2 s, \ldots, \Delta_n s$ respectively. Suppose that $f = F \circ \pi$ is a bounded cylinder function on $W(K)$. Define $\tilde{F} : K^n \to \mathbb{R}$ so that
\[
F(x_1, x_2, x_3, \ldots, x_n) = \tilde{F}(x_1, x_2x_1^{-1}, x_3x_2^{-1}, \ldots, x_nx_{n-1}^{-1})
\]
for all $x_i \in K$. Then
\[
f(\Sigma(t, \cdot)) = \tilde{F}(U_1(t), U_2(t), \ldots, U_n(t))
\]
and therefore
\[
\nu_t(f) = \mathbb{E}f(\Sigma(t, \cdot)) = \mathbb{E}\tilde{F}(U_1(t), U_2(t), \ldots, U_n(t))
\]
(4.6)
\[
= \int_{K^n} \tilde{F}(x_1, \ldots, x_n) \prod_{i=n}^n p^K_{\Delta_i s}(x_i) dx_i.
\]
Let $x_0 := e$. Using the invariance of Haar measure, make the translations
\[
x_2 \to x_2x_1^{-1} \text{ then}
\]
\[
x_3 \to x_3x_2^{-1} \text{ then}
\]
\[
\vdots
\]
\[
x_n \to x_nx_{n-1}^{-1}
\]
in the last integral of Eq. (4.6) to find
\[
\nu_t(f) = \int_{K^n} \tilde{F}(x_1, x_2x_1^{-1}, \ldots, x_nx_{n-1}^{-1}) \prod_{i=n}^n p^K_{\Delta_i s}(x_i) dx_i
\]
\[
= \int_{K^n} F(x_1, x_2, x_3, \ldots, x_n) \prod_{i=n}^n p^K_{\Delta_i s}(x_i) dx_i
\]
(4.7) \[ = \int_{K^n} F(x_1, x_2, x_3, \ldots, x_n) \prod_{i=n}^{n} p_{K}^K (x_{i-1}^{-1} x_i) dx_i \]

wherein the last equality we have use the fact that \( p_{K}^K (\cdot) \) is a class function, see Remark 2.10.

Comparing Eq. (4.7) with Eq. (2.8), shows that \( \nu_t (f) = \mu_t (f) \) for all bounded cylinder functions \( f \) on \( W(K) \) which implies that \( \nu_t = \mu_t \) by Lemma 2.7.

5. Proof of the Airault-Malliavin Theorem 2.18 This subsection is devoted to the proof of Theorem 2.18. We will need some, mostly well known, preliminary results regarding integration by parts on \((W(K), \mu_t)\): These results will be gathered in the next subsection.

5.1. Integration by parts and strong differentiability The key result here for the remainder of the paper is Corollary 5.6. The reader may skip this subsection if she/he is willing to accept Corollary 5.6 below.

DEFINITION 5.1. Let \( L_{1}^{-} (W(K), \mu_t) = \cap_{1 \leq p < \infty} L_{p}^p (W(K), \mu_t) \) and \( h \in H \). A function \( f \in L_{1}^{-} (W(K), \mu_t) \) is said to be strongly \( h \) differentiable provided there is a function \( g \in L_{1}^{-} (W(K), \mu_t) \) such that

\[
g = L_{p}^p (\mu_t) - \lim_{\epsilon \to 0} \frac{f(\sigma e^{\epsilon h}) - f(\sigma)}{\epsilon}
\]

for all \( 1 \leq p < \infty \). We will denote the function \( g \), if it exists, by \( \partial_h f \).

Cylinder functions are strongly \( h \)-differentiable for all \( h \in H \) and \( \partial_h f \) is given by Eq. (3.4). Another example is given in Lemma 5.5 below.

DEFINITION 5.2. An element \( k \in W(K) \) is a finite energy path if

\[
k' (s) \text{ exists } ds\text{-a.s. and } \int_0^1 \left| k^{-1}(s)k'(s) \right|^2 ds < \infty.
\]

Letting \( k \in W(K) \) be a finite energy path and \( b_s \) being as in Eq. (2.17), then for \( \mu_t \) - a.e. \( \sigma \in W(K) \),

\[
b_s (\sigma k) = \int_0^s (\sigma(r) k(r))^{-1} \delta (\sigma k) (r)
= \int_0^s k^{-1}(r)\sigma^{-1}(r) \left[ \delta(\sigma(r) k(r) + \sigma(r) k'(r)) \right] dr
= \int_0^s Ad_{k^{-1}(r)} db_s (\sigma) + \int_0^s k^{-1}(r) k'(r) dr.
\]

(5.1) Since \( Ad_{k^{-1}(r)} \) is orthogonal on \( t \), Lévy’s characterization of Brownian motion shows that \( B_s := \int_0^s Ad_{k^{-1}(r)} db_r \) on \( (W(K), \mu_t) \) is still a Brownian motion with variance \( t \). This observation and the Cameron-Martin theorem is essentially the proof of the following quasi invariance theorem of Albeverio and Høegh-Krohn, see [4], [30], and [29].
THEOREM 5.3 (Albeverio & Hoegh-Krohn). Let \( k \) be a finite-energy path on \( K \) and \( f : W(K) \to \mathbb{R} \) be a bounded measurable function. Then

\[
\int_{W(K)} f(\sigma) d\mu_\sigma(\sigma) = \int_{W(K)} f(\sigma k) J_k(\sigma) d\mu_\sigma(\sigma),
\]

where

\[
J_k := \exp\left(-\frac{1}{t} \int_0^1 \langle k'(s) k^{-1}(s), dB_s \rangle - \frac{1}{2t} \int_0^1 |k'(s) k^{-1}(s)|^2 ds \right).
\]

PROOF OF THEOREM 5.3. Let \( h(s) := \int_0^s k^{-1}(r) k'(r) dr \), \( B_s := \int_0^s Ad_{k^{-1}(r)} dB_r \), and \( \tilde{f} \) be a measurable function on \( C([0,1] \to \mathfrak{g}) \) such that \( \tilde{f}(b(\sigma)) = f(\sigma) \) for \( \mu_t \)-a.e. \( \sigma \). Using the \( Ad_K \)-invariance of the inner product \( \langle \cdot, \cdot \rangle \) on \( \mathfrak{g} \), we have

\[
\int_0^1 |k'(r) k^{-1}(r)|^2 dr = \int_0^1 |k^{-1}(r) k'(r)|^2 dr
\]

and

\[
\int_0^1 \langle k'(r) k^{-1}(r), dB_r \rangle = \int_0^1 \langle k'(r) k^{-1}(r), Ad_{k^{-1}(r)} dB_r \rangle
\]

\[
= \int_0^1 \langle Ad_{k^{-1}(r)} k'(r) k^{-1}(r), dB_r \rangle
\]

\[
= \int_0^1 \langle k^{-1}(r) k'(r), dB_r \rangle.
\]

Combining these equations show that \( J_k \) may be written as

\[
J_k := \exp\left(-\frac{1}{t} \int_0^1 \langle h', dB_r \rangle - \frac{1}{2t} \int_0^1 |h'(r)|^2 dr \right).
\]

By Eq. (5.1),

\[
\int_{W(K)} f(\sigma k) J_k(\sigma) d\mu_\sigma(\sigma) = \int_{W(K)} \tilde{f}(b(\sigma k)) J_k(\sigma) d\mu_\sigma(\sigma)
\]

\[
= \int_{W(K)} \tilde{f}(B(\sigma) + h) J_k(\sigma) d\mu_\sigma(\sigma)
\]

\[
= \int_{W(K)} \tilde{f}(B(\sigma)) d\mu_\sigma(\sigma)
\]

(5.5)

wherein the last equality we have used the Cameron-Martin (or Girsanov’s) theorem. Since \( B \) and \( b \) have the same laws, being \( \mathfrak{g} \)-valued Brownian motions with variance \( t \),

\[
\int_{W(K)} \tilde{f}(B(\sigma)) d\mu_\sigma(\sigma) = \int_{W(K)} \tilde{f}(b(\sigma)) d\mu_\sigma(\sigma) = \int_{W(K)} f(\sigma) d\mu_\sigma(\sigma).
\]

(5.6)

Combining Eqs. (5.4), (5.5) and (5.6) proves the theorem.
Corollary 5.4. Let \( h \in H(t) \) and suppose that \( f \) and \( g \) are strongly \( h \)-differentiable, then
\[
\mu(g \partial_h f) = \mu(( -\partial_h g + j_h g) f)
\]
where
\[
j_h := \frac{1}{t} \int_0^1 \langle h'(s), db_s \rangle.
\]

This corollary has been proved in the more general context of Wiener measure on a Riemannian manifold in Driver [12].

Proof of Corollary 5.4. Let \( k = e^{th} \) and replace \( f \) by \( fg \) in Eq. (5.2) of Theorem 5.3 to find
\[
\mu(fg) = \int_{W(K)} f(\sigma e^{th}) g(\sigma e^{th}) J_{e^{th}}(\sigma) d\mu(\sigma).
\]
Differentiate this equation in \( \epsilon \) implies
\[
0 = \mu \left( \partial_h f \cdot g + f \partial_h g + f g \frac{d}{dt} \langle J_{e^{th}} \rangle \right)
\]
which proves the corollary provided that
\[
\frac{d}{d\epsilon} \langle J_{e^{th}} \rangle = -j_h \text{ in } L^p(\mu) \text{ for all } p \in [1, \infty).
\]
We will not carry out the convergence details here which are fairly routine. The interested reader may refer to Gross [21] or Section 9 in [12]. However, let us check "algebraically" that the formula in Eq. (5.8) is correct. Computing \( \frac{d}{d\epsilon} \langle J_{e^{th}} \rangle \) gives
\[
\frac{d}{d\epsilon} \langle J_{e^{th}} \rangle = \frac{d}{d\epsilon} \langle J_{e^{th}} \rangle = \frac{d}{d\epsilon} \left( \frac{d}{dr} e^{\epsilon h(r)} e^{-\epsilon h(r)} \right) dB_r
\]
which proves the corollary provided that
\[
\frac{d}{d\epsilon} \langle J_{e^{th}} \rangle = -j_h \text{ in } L^p(\mu) \text{ for all } p \in [1, \infty).
\]
Lemma 5.5. For each $h \in H$, the function $j_h$ is strongly $h$ differentiable and

$$
\partial_h j_h = \frac{1}{t} \int_0^1 \langle a d_{h(r)} h'(r), d b_r(\sigma) \rangle + \frac{1}{t} \int_0^1 |h'(r)|^2 dr.
$$

Proof of Lemma 5.5. According to Eq. (5.1),

$$
j_h(\sigma e^{ch}) = \frac{1}{t} \int_0^1 \left( h'(r), A d_{e^{ch(r)}} d b_r(\sigma) + e^{-ch(r)} \frac{d}{dr} e^{ch(r)} dr \right)$$

$$= \frac{1}{t} \int_0^1 \left( h'(r), d b_r(\sigma) \right) + \frac{1}{t} \int_0^1 \left( h'(r), e^{-ch(r)} \frac{d}{dr} e^{ch(r)} dr \right).
$$

Therefore, again ignoring convergence questions,

$$
\partial_h j_h(\sigma) = \frac{d}{d\sigma} \int_0^1 \langle a d_{h(r)} h'(r), d b_r(\sigma) \rangle + \frac{1}{t} \int_0^1 |h'(r)|^2 dr.
$$

Here the convergence questions are even easier since we only have jointly Gaussian random variables to contend with and $L^2$ convergence of Gaussian random variables implies $L^p$ convergence for $p < \infty$. The reader may find more details in Section 4 in Gross [20].

The following Corollary is a key ingredient in our proof of the Airault–Malliavin Theorem 2.18.

Corollary 5.6. Let $f$ be a smooth cylinder function (see Definition 2.4) and $h \in H(\mathbb{R})$ such that the Lie bracket $[h(s), h'(s)] = 0$ for a.e. $s$. Then

$$
\mu_t(\partial_h^2 f) = \mu_t \left( j_h - \frac{1}{t} \int_0^1 |h'(r)|^2 dr \right) f,$$

where $j_h$, as in Eq. (5.8)

Proof of Corollary 5.6. Two applications of Corollary 5.4 gives

$$
\mu_t(\partial_h^2 f) = \mu_t(j_h \partial_h f) = \mu_t((-\partial_h j_h + j_h^2) f)
$$

which combined with Eq. (5.9) of Lemma 5.5 proves the Corollary.

5.2. Proof of Theorem 2.18

Proof of Theorem 2.18. Let $f = F \circ \pi_\alpha$ be a cylinder function on $L(K)$ (see Definition 2.4) and let $\alpha \in (s_n, 1)$. (We will eventually let $\alpha \to 1$.) Recall the definition of pinned Wiener measure $\mu^K_\alpha$ (see Definition 2.11) says that $\mu^K_\alpha(f) = \mu_\alpha(f \eta)$ where $\eta := p^K_{1}(1-\alpha)(\pi_\alpha)/p^K_{1}(e)$. Therefore, by Corollary 4.1,

$$
\partial_t \mu^K_\alpha(f) = \partial_t \mu_\alpha(f \eta)
$$

$$= \mu_\alpha(f \partial_\eta \eta) + \frac{1}{2} \mu_\alpha(\Delta_{W(K)}(f \eta))
$$

$$= I_\alpha + J_\alpha.
$$
Now
\[ I_\alpha = \mu_t(f \partial_t \eta_t) \]
\[ = \frac{1}{2p^K_p(e)}(1 - \alpha)\mu_t \left( f \Delta K p^K_{1(1 - \alpha)}(\pi_\alpha) \right) - \partial_t \ln p^K_p(e)\mu_t^0(f). \]

By Eq. (2.8),
\[ \mu_t(f \Delta K p^K_{1(1 - \alpha)}(\pi_\alpha)) = \int_K G(\alpha, x) \Delta K p^K_{1(1 - \alpha)}(x)dx \]
\[ = \int_K \Delta K G(\alpha, x)p^K_{1(1 - \alpha)}(x)dx \]
where
\[ G(\alpha, x) := \int_{K^n} F(x_1, \cdots, x_n) p^K_{1(1 - \alpha)}(x_1) \prod_{i=1}^n p^K_{\Delta_i s}(x_i^{-1}x_i) dx_i. \]

From this expression we see that \( \Delta K G(\alpha, x) \) remains bounded as \( \alpha \to 1 \), so that letting \( \alpha \to 1 \) in Eq. (5.11) gives
\[ \lim_{\alpha \to 1} I_\alpha = -\mu^K_t(f \partial_t \log p^K_p(e)). \]

We proceed to work on the second term, \( J_\alpha \), in Eq. (5.10). Let \( P_n \) be the partition of \([0, 1]\),
\[ P_n = \{ 0 = s_0 < s_1 < s_2 < \ldots < s_n < \alpha < 1 \}, \]
and set \( s_{n+1} = \alpha \). Define \( G^K_0(s, t) = (s \wedge t - \alpha^{-1}st) \) so that
\[ G(s, t) = s \wedge t = G^K_0(s, t) + \alpha^{-1}st. \]

Let \( \eta_t(x_1, x_2, \ldots, x_{n+1}) = p^K_{1(1 - \alpha)}(x_{n+1})/p^K_p(e) \) and by abuse of notation use \( F \) again to denote the function \( (x_1, x_2, \ldots, x_{n+1}) \in K^{n+1} \to F(x_1, x_2, \ldots, x_n) \). Then by Eqs. (3.10) and (3.8) applied to the partition \( P_n \),
\[ \Delta W_K(f \eta_t) = L_{P_n}(F \eta_t) \circ \eta_{P_n} \]
\[ = \sum_{i,j=1}^{n+1} \sum_{A \in \ell_0} G^K_0(s_i, s_j) A^{(i)} A^{(j)}(F \eta_t) \circ \eta_{P_n} \]
\[ + \sum_{i,j=1}^{n+1} \sum_{A \in \ell_0} \alpha^{-1}s_i s_j A^{(i)} A^{(j)}(F \eta_t) \circ \eta_{P_n} \]
\[ = S_\alpha + T_\alpha. \]

Now for \( A \in \ell \), let
\[ h^K_\alpha(s) := \alpha^{-1/2}(s \wedge \alpha)A. \]
Then by Eq. (3.5),
\[ T_\alpha = \sum_{A \in \ell_0} \partial^2 h^K_\alpha(f \eta_t). \]
For the $S_\alpha$ term in Eq. (5.13), notice that by construction $G_0^\alpha(s,t) = 0$ if $s$ or $t$ is in $\{0, \alpha\}$. Therefore $G_0^\alpha(s_i, s_j) = 0$ if $i$ or $j = n + 1$ (i.e. $s_i$ or $s_j$ is $\alpha$) so that

$$S_\alpha = \sum_{i,j=1}^n \sum_{A \in \Theta_0} G_0^\alpha(s_i, s_j) A^{(i)} A^{(j)}(F \eta) \circ \pi_\alpha$$

$$= \tilde{\eta} \circ \pi_\alpha \cdot \sum_{i,j=1}^n \sum_{A \in \Theta_0} G_0^\alpha(s_i, s_j) (A^{(i)} A^{(j)} F) \circ \pi_\alpha. \tag{5.15}$$

Taking the $\mu_t$ expectation of Eq. (5.13) and making use of Eq. (5.14) and (5.15) shows that $J_\alpha$ from Eq. (5.10) satisfies

$$J_\alpha = \frac{1}{2} \frac{\mu_t^0}{\mu_t} \left( \sum_{i,j=1}^n \sum_{A \in \Theta_0} G_0^\alpha(s_i, s_j) (A^{(i)} A^{(j)} F) \circ \pi_\alpha \right)$$

$$+ \frac{1}{2} \mu_t \left( \sum_{A \in \Theta_0} \partial^2 h_\alpha^F(f \eta) \right)$$

$$= J_\alpha^{(1)} + J_\alpha^{(2)}.$$

Since $G_0^\alpha \to G_0$ as $\alpha \to 1$,

$$\lim_{\alpha \to 1} J_\alpha^{(1)} = \frac{1}{2} \frac{\mu_t^0}{\mu_t} \left( \sum_{i,j=1}^n \sum_{A \in \Theta_0} G_0(s_i, s_j) (A^{(i)} A^{(j)} F) \circ \pi_\alpha \right)$$

$$= \frac{1}{2} \frac{\mu_t^0}{\mu_t} \left( \sum_{i,j=1}^{n+1} \sum_{A \in \Theta_0} G_0(s_i, s_j) (A^{(i)} A^{(j)} F) \circ \pi_\alpha \right)$$

$$= \frac{1}{2} \mu_t^0 (\Delta L(K)f), \tag{5.16}$$

where we have used Eq. (3.10) for the last equality.

By Corollary 5.6,

$$2J_\alpha^{(2)} = \sum_{A \in \Theta_0} \mu_t \left( \left[ \frac{j^2 h_\alpha^\lambda}{\lambda} - \frac{1}{t} \int_0^1 \left| \frac{d}{dr} h_\alpha^\lambda(r) \right|^2 dr \right] f \eta \right)$$

$$= \sum_{A \in \Theta_0} \mu_t \left( \left[ \frac{j^2 h_\lambda^\lambda}{\lambda} - \frac{1}{t} \int_0^1 \left| \frac{d}{dr} h_\lambda^\lambda(r) \right|^2 dr \right] f \right),$$

where (by Eq. (5.8))

$$j h_\alpha = \frac{1}{\sqrt{\alpha t}} \int_0^\alpha \langle A, db_s \rangle = \frac{1}{\sqrt{\alpha t}} \int_0^\alpha \langle A, db_s \rangle = \frac{1}{\sqrt{\alpha t}} \langle A, b_0 \rangle.$$

Using these facts and

$$\int_0^1 \frac{d}{dr} h_\lambda^\lambda(r)^2 dr = \frac{1}{t} \int_0^\alpha |A|^2 dr = |A|^2.$$
we see that
\[ J^{(2)}_a = \frac{1}{2} \mu^0_t \left( \frac{1}{\alpha t^2} |b_0|^2 - \frac{1}{t} \dim \mathfrak{t} \right) \]
and hence by Remark 2.17,
\[ \lim_{a \to 1} J^{(2)}_a = \frac{1}{2} \mu^0_t \left( \frac{1}{\alpha t^2} |b_0|^2 - \frac{1}{t} \dim \mathfrak{t} \right). \]
Assembling Eqs. (5.16) and (5.17) shows that
\[ \lim_{a \to 1} J_a = \frac{1}{2} \mu^0_t \left( \frac{1}{\alpha t^2} |b_0|^2 - \frac{1}{t} \dim \mathfrak{t} \right). \]
Combining Eqs. (5.10), (5.12) and (5.18) proves the Theorem.

**Corollary 5.7.** Suppose that \( u : \mathcal{L}(K) \to \mathbb{R} \) is a smooth cylinder function and let
\[ G(t, \sigma) = \int_{\mathcal{L}(K)} u(\sigma^{-1}) d\mu^0_t(\gamma), \]
then
\[ \partial_t G(t, \sigma) = \frac{1}{2} \Delta_{\mathcal{L}(K)} G(t, \sigma) + \int_{\mathcal{L}(K)} V_0(\gamma) u(\sigma^{-1}) d\mu^0_t(\gamma). \]

**Proof of Corollary 5.7.** As in the proof of Corollary 3.11, let \( u_\sigma : \mathcal{L}(K) \to \mathbb{R} \) be the cylinder function defined by \( u_\sigma(\gamma) = u(\sigma^{-1}) \). By the Airault Malliavin Theorem 2.18,
\[ \partial_t G(t, \sigma) = \int_{\mathcal{L}(K)} \left[ \frac{1}{2} \Delta_{\mathcal{L}(K)} u_\sigma(\gamma) + V_0(\gamma) u_\sigma(\gamma) \right] d\mu^0_t(\gamma). \]
Using the same method of proof used for Corollary 3.11, we see that this equation is the same as Eq. (5.20).

6. Absolute continuity of heat kernel with respect to pinned Wiener measure In this section we will prove the main Theorem 2.16. We will first need a couple of preliminary results.

**Lemma 6.1 (Asymptotic properties of heat Kernels on K).** The heat kernel, \( \mu^K_t \), on \( K \) has the following properties:

1. \( \lim_{t \to 0} (2\pi t)^{-\frac{1}{2} \dim \mathfrak{t}} \mu^K_t(e) = 1 \).

2. For every \( T < \infty \), there is a constant \( M_T < \infty \) such that
\[ \mu^K_t(x) \leq M_T t^{-\frac{1}{2} \dim \mathfrak{t}} e^{-\frac{1}{2} d^2(x, e)} \]
for all \( x \in K \) and \( 0 < t \leq T \)
where \( d(x, y) \) is the distance associated to the bi-invariant Riemannian metric on \( K \) which agrees with \( \langle \cdot, \cdot \rangle \) at \( e \in K \).
Proof of Lemma 6.1. These are standard properties of heat kernels. For item 1., see Theorem 2.30 of [6]. See also [26]. For the second item see, for example, Theorem IX.1.2 in [33]. To apply this theorem, use the fact that $K$ is compact so the modular function is constant. It is also necessary to note that the time parameter in [33] is twice our time parameter $t$.

**Lemma 6.2** ($\mu_t^0 \to \delta_x$ as $t \to 0$). Let $f : \mathcal{L}(K) \to \mathbb{R}$ be a continuous cylinder function, then

$$\lim_{t \to 0^+} \mu_t^0(f) = f(e),$$

where $e$ denotes the identity loop in $\mathcal{L}(K)$, see Notation 1.2.

Proof of Lemma 6.2. This result can be proved in a number of ways. For example one could use the Kolmogorov’s continuity criteria to show that $\mu_t^0$ concentrates near the identity loop as $t \to 0$. See the argument in the proof of Item 1 of Theorem 2.3 in [13]. Rather than carry this out in full detail, we will only prove what we need.

Let $\mathcal{P}$ be a partition of $[0, 1]$ as in Eq. (2.3), $f = F \circ \pi_{\mathcal{P}}$ and $\rho^\mathcal{P} : (0, \infty) \times K^n \to (0, \infty)$ be as in Eq. (2.10). By Lemma 6.1, there is a constant $M < 1$ such that

$$\rho^\mathcal{P}(t, x) \leq M t^{\frac{1}{2} \dim K} \prod_{i=1}^{n+1} (t \Delta_i s)^{-\frac{\dim K}{2}} \exp \left( -\frac{1}{4t \Delta_i s} d^2(e, x_{i-1}^1, x_i) \right)$$

for all $t \in (0, 1]$, where $x = (x_1, \cdots, x_n)$, $\Delta_i s = s_i - s_{i-1}$, and $x_0 = x_{n+1} = e \in K$. By the left invariance of the Riemannian metric on $K$, $d(x, y) = d(e, x^{-1}y)$, so the previous inequality may be written as

$$\rho^\mathcal{P}(t, x) \leq M_\mathcal{P} t^{-\frac{1}{2} \dim K} \exp \left( -\frac{1}{4t} \sum_{i=1}^{n+1} \frac{d^2(x_{i-1}, x_i)}{\Delta_i s} \right)$$

where $M_\mathcal{P} := M \prod_{i=1}^{n+1} (\Delta_i s)^{-\frac{\dim K}{2}}$. Now let $\delta > 0$ be given, and suppose that $d(e, x_i) \geq \delta$ for some $i \in \{1, 2, \ldots, n\}$, then by the triangle inequality and the Cauchy–Schwarz inequality,

$$\delta^2 \leq d^2(e, x_i) \leq \left( \sum_{j=1}^{i} d(x_{j-1}, x_j) \right)^2 \leq \left( \sum_{i=1}^{n+1} \frac{d(x_{j-1}, x_j)}{\sqrt{\Delta_j s}} \sqrt{\Delta_j s} \right)^2 \leq \sum_{i=1}^{n+1} \Delta_i s \cdot \sum_{i=1}^{n+1} \frac{d^2(x_{i-1}, x_i)}{\Delta_i s} \Delta_i s \sum_{i=1}^{n+1} \frac{d^2(x_{i-1}, x_i)}{\Delta_i s}$$

Combining this estimate with Eq. (6.2) implies

$$\rho^\mathcal{P}(t, x) \leq M_\mathcal{P} t^{-\frac{1}{2} \dim K} \exp \left( -\frac{1}{4t} |x|^2 \right)$$
where

\[ |x| := \max\{d(e, x_i) : i = 1, 2, \ldots, d\}. \]

Therefore \( \rho^\beta(t, \cdot) \) satisfies:

1. \( \rho^\beta(t, x) > 0 \).
2. \( \int_{K^n} \rho^\beta(t, x)dx = 1 \) where \( dx \) is Haar measure on \( K^n \).
3. For any \( \delta > 0, \rho^\beta(t, x) \to 0 \) uniformly in \( x \in K^n \) with \( |x| \geq \delta \).

It is now routine to show, using these three properties, that

\[
\lim_{t \to 0} \int_{K^n} F(x) \rho^\beta(t, x)dx = F(e, e, \ldots, e)
\]

which is equivalent to Eq. (6.1).

6.1. Proof of Theorem 2.16

**Proof of Theorem 2.16.** Let \( u \) be a smooth non-negative cylinder function on \( L(K) \) and let \( C_t \) be as in Eq. (2.16). Notice that \( \frac{\partial}{\partial t} C_t = c_t \) (\( c_t \) is defined in Eq. (2.19) of the Airault – Malliavin Theorem 2.18) and because of Lemma 6.1, \( \lim_{t \to 0} C_t = 0 \). Define

\[
H(t, \sigma) = \int_{\mathcal{L}(K)} u(\sigma \gamma^{-1})d\mu_t^0(\gamma), \quad \text{and}
\]

\[
F(t, \sigma) = e^{C_t} \int_{\mathcal{L}(K)} u(\sigma \gamma^{-1})d\mu_t^0(\gamma),
\]

then by Corollary 3.11

\[
\partial_t H(t, \sigma) = \frac{1}{2} \Delta_{\mathcal{L}(K)} H(t, \sigma) \quad \text{and} \quad \lim_{t \to 0} H(t, \sigma) = u(\sigma)
\]

and by Corollary 5.7

\[
\partial_t F(t, \sigma) = \frac{1}{2} \Delta_{\mathcal{L}(K)} F(t, \sigma) + e^{C_t} \int_{\mathcal{L}(K)} (V_t(\gamma) + c_t)u(\sigma \gamma^{-1})d\mu_t^0(\gamma)
\]

\[
\geq \frac{1}{2} \Delta_{\mathcal{L}(K)} F(t, \sigma).
\]

Combining this with Lemma 6.2, shows that

\[
\partial_t F(t, \sigma) \geq \frac{1}{2} \Delta_{\mathcal{L}(K)} F(t, \sigma) \quad \text{and} \quad \lim_{t \to 0} F(t, \sigma) = u(\sigma).
\]

The idea now is to use Eqs. (6.4), (6.5) and the maximum principle to conclude that

\[
F(t, \sigma) \geq H(t, \sigma) \quad \text{for all} \quad 0 \leq t < \infty \quad \text{and} \quad \sigma \in \mathcal{L}(K).
\]
We will postpone the full justification of Eq. (6.6) to Lemma 6.3 below.

Writing out Eq. (6.6) when $\sigma$ is the constant loop $e$, shows that

$$
\int_{\mathcal{L}(K)} u(\gamma^{-1}) du^0_\sigma(\gamma) \leq e^{C_1} \int_{\mathcal{L}(K)} u(\gamma^{-1}) d\mu^0_\sigma(\gamma)
$$

for all non-negative smooth cylinder functions $u$. Replacing $u$ by the cylinder function $u(\gamma) = u(\gamma^{-1})$ then implies that

$$
\int_{\mathcal{L}(K)} u(\gamma) du^0_\sigma(\gamma) \leq e^{C_1} \int_{\mathcal{L}(K)} u(\gamma) d\mu^0_\sigma(\gamma)
$$

for all non-negative smooth cylinder functions $u$.

Since, by Lemma 2.7, bounded smooth cylinder functions are dense in $L^2(\mathcal{L}(K), \mathcal{F}, \mu^0 + \nu^0)$, by passing to the limit, we may conclude that Eq. (6.7) is valid for all bounded non-negative measurable functions $u$. By taking $u$ to be characteristic functions and using the Radon-Nikodym theorem, Eq. (6.7) implies that $\nu^0$ is absolutely continuous relative to $\mu^0$. Letting $Z_t := dv^0_\sigma/d\mu^0_\sigma$ we may conclude from Eq. (6.7) that

$$
\int_{\mathcal{L}(K)} u \cdot (Z_t - e^{C_1}) d\mu^0_\sigma \leq 0
$$

for all bounded measurable functions $u$ and hence that $Z_t - e^{C_1} \leq 0$.

**Lemma 6.3.** Keeping the same notation as above, Eq. (6.6) is valid.

**Proof of Lemma 6.3.** In order to justify the use of the maximum principle to prove Eq. (6.6), write $u = U \circ \pi_\mathbb{P}$, where $\mathbb{P}$ is a partition as in Eq. (2.3) and $U : K^n \to [0, \infty)$ is a smooth function. Then

$$
H(t, \sigma) = \int_{\mathcal{L}(K)} u(\sigma \gamma^{-1}) d\mu^0_\sigma(\gamma) = \int_{\mathcal{L}(K)} U(\pi_\mathbb{P}(\sigma) \pi_\mathbb{P}(\gamma)^{-1}) d\mu^0_\sigma(\gamma)
$$

(6.8)

$$
= H_{\mathbb{P}}(t, \pi_\mathbb{P}(\sigma)),
$$

where for $x \in K^n$

$$
H_{\mathbb{P}}(t, x) = \int_{\mathcal{L}(K)} U(x \pi_\mathbb{P}(\gamma)^{-1}) d\mu^0_\sigma(\gamma)
$$

(6.9)

$$
= \int_{K^n} U(xy^{-1}) \bar{p}_{\mathbb{P}}(y) dy,
$$

and $\bar{p}_{\mathbb{P}}(y) dy = Law(\pi_\mathbb{P}(\Sigma(t, \cdot)))$. By the proof of Proposition 3.9, $\pi_\mathbb{P}(\Sigma(t, \cdot))$ is a diffusion on $K^n$ with elliptic generator $L^0_\mathbb{P}$ defined in Eq. (3.9). Thus $\bar{p}_{\mathbb{P}}(y)$ is the smooth heat kernel for the operator $e^{tL^0_\mathbb{P}}$. This shows that $H_{\mathbb{P}}(t, x)$ is smooth on $(0, \infty) \times K^n$. Using this information, Eq. (6.4) may be recast as the finite dimensional statement

$$
\partial_t H_{\mathbb{P}}(t, x) = \frac{1}{2} L^0_\mathbb{P} H_{\mathbb{P}}(t, x) \text{ and } \lim_{t \to 0} H_{\mathbb{P}}(t, x) = U(x).
$$

(6.10)
Similarly
\[ F(t, \sigma) = e^{Ct} \int_{\mathcal{L}(K)} u(\sigma \gamma^{-1}) \, d\mu_t^0(\gamma) = e^{Ct} \int_{\mathcal{L}(K)} U(\pi_{\mathcal{P}}(\sigma) \pi_{\mathcal{P}}(\gamma)^{-1}) \, d\mu_t^0(\gamma) \]
\[ = F_{\mathcal{P}}(t, \pi_{\mathcal{P}}(\sigma)), \]
where for \( x \in K^n \)
\[ F_{\mathcal{P}}(t, x) = e^{Ct} \int_{\mathcal{L}(K)} U(x \pi_{\mathcal{P}}(\gamma)^{-1}) \, d\mu_t^0(\gamma) \]
\[ = e^{Ct} \int_{K^n} U(xy^{-1}) \rho_{\mathcal{P}}(t, y) \, dy \]
where \( \rho_{\mathcal{P}} : (0, \infty) \times K^n \rightarrow (0, \infty) \) is the smooth function defined in Eq. (2.10) of Definition 2.11. This shows that \( F_{\mathcal{P}}(t, x) \) is smooth on \((0, \infty) \times K^n\). Using this information, Eq. (6.4) may be recast as the finite dimensional statement
\[ \partial_t F_{\mathcal{P}}(t, x) \geq \frac{1}{2} L_{\mathcal{P}} F_{\mathcal{P}}(t, x) \text{ and } \lim_{t \to 0} F_{\mathcal{P}}(t, x) = U(x). \]

Now there is no problem in applying the maximum principle on \( K^n \), using Eqs. (6.10) and (6.11), to conclude that
\[ F_{\mathcal{P}}(t, x) \geq H_{\mathcal{P}}(t, x) \text{ for all } 0 \leq t < \infty \text{ and } x \in K^n. \]
This finishes the proof since this last assertion is equivalent (6.6).

7. The \( K = \mathbb{R}^d \) and \( S^1 \) cases
In this section, we will work out the explicit relationship between \( \mu_t^0 \) and \( \nu_t^0 \) in the case that \( K \) is the abelian Lie group \( \mathbb{R}^d \) or \( S^1 \).

7.1. The \( K = \mathbb{R}^d \) case
Let \( K \) be the Lie group \( \mathbb{R}^d \) with group operation being addition. The Lie algebra of \( \mathbb{R}^d \) is \( \mathfrak{k} = \mathbb{R}^d \) with the trivial Lie bracket, \([a, b] = 0\) for all \( a, b \in \mathbb{R}^d \). Although \( \mathbb{R}^d \) is not compact and is not being represented as a matrix group, the theory above easily extends to this case. There is one notational point to take care of now. Namely, the matrix expression of the form \( g^{-1} \delta g \) must now be interpreted as \( L_{g^{-1}} \delta g = \delta g \). We will assume that \( \langle a, b \rangle = a \cdot b \) is the usual dot product, although any inner product would work.

Lemma 7.1. On the loop space of \( \mathbb{R}^d, \mathcal{L}(\mathbb{R}^d) \), the heat kernel measures \( \nu_t^0 \) and the pinned Wiener measures, \( \mu_t^0 \), are the same.

Proof of Lemma 7.1. The process \( \Sigma^0(t, s) \) in Theorem 2.13 and the process \( g \) in Eq. (2.3) of Definition 2.8 are explicitly given by \( \Sigma^0(t, s) = \chi(t, s) \) and \( g_s = \beta(t, s) \) respectively. Since \( g_s = \beta(t, s) \) is a standard Brownian motion with variance \( t \), the pinned Wiener measure \( \mu_t^0 = \text{Law}(g, |g| = 0) \) is the law of an \( \mathbb{R}^d \)-valued Brownian bridge with variance \( t \). But \( s \to \chi(t, s) \) is a Brownian bridge with variance \( t \) (see Remark 2.12), so that \( \mu_t^0 = \text{Law}(\chi(t, \cdot)) = \text{Law}(\Sigma^0(t, \cdot)) = \nu_t^0 \).
7.2. The $K = S^1$ case Let $K = S^1 = \{ z \in \mathbb{C} : |z| = 1 \}$. The Lie algebra of $K$ is $\mathfrak{k} = i\mathbb{R}$ with the trivial Lie bracket. We will identify with $\mathfrak{k} = i\mathbb{R}$ with $\mathbb{R}$, putting in the $i$ explicitly when needed. Let $\langle a, b \rangle = ab$ for $a, b \in \mathbb{R} \cong i\mathbb{R} = \mathfrak{k}$.

Remark 7.2. Let $p_\ell(x) = p_\ell^\mathbb{R}(x) = (2\pi)\ell^{-1/2} \exp \left( -\frac{1}{2\beta^2} x^2 \right)$ be the heat kernel on $\mathbb{R}$, and $q_\ell(z) = \frac{1}{2\pi^2} \hat{p}_\ell(z)$ denote the heat kernel on $S^1$ relative to the un-normalized Haar measure $\cdot d\theta$, i.e. for $f : S^1 \to \mathbb{R}$,

$$\int_{S^1} f d\theta := \int_0^{2\pi} f(e^{i\theta}) d\theta.$$  

The well known relationship between $q_\ell$ and $p_\ell$ is

$$q_\ell(e^{i\theta}) = \sum_{n=-\infty}^{\infty} p_\ell(\theta - 2\pi n) \quad \text{for} \quad \theta \in \mathbb{R}. \quad (7.1)$$

To check Eq. (7.1), suppose that $f_0 : S^1 \to \mathbb{R}$ is a continuous function. Then

$$f(t, z) = \int_0^{2\pi} f_0(ze^{-i\alpha}) q_\ell(e^{i\alpha}) d\alpha$$

solves the heat equation on $S^1$ which is equivalent to saying that $F(t, \alpha) := f(t, e^{i\alpha})$ solves the heat equation on $\mathbb{R}$. Since $F$ is a bounded solution to the heat equation it is given by

$$\int_0^{2\pi} f_0(ze^{-i\alpha}) q_\ell(e^{i\alpha}) d\alpha = F(t, \theta) = \int_{-\infty}^{\infty} F(0, \theta - \alpha)p_\ell(\alpha) d\alpha$$

$$= \int_{-\infty}^{\infty} f_0(ze^{-i\alpha}) p_\ell(\alpha) d\alpha$$

$$= \sum_{n=-\infty}^{\infty} \int_0^{2\pi} f_0(ze^{-i(\alpha - 2\pi n)}) p_\ell(\alpha - 2\pi n) d\alpha$$

$$= \sum_{n=-\infty}^{\infty} \int_0^{2\pi} f_0(ze^{-i\alpha}) p_\ell(\alpha - 2\pi n) d\alpha$$

where $z = e^{i\theta}$. This equation, holding for all continuous $f_0 : S^1 \to \mathbb{R}$, proves Eq. (7.1).

Definition 7.3. For $n \in \mathbb{Z}$, let $h_n(s) := 2\pi ns$,

$$z_n(s) = e^{2\pi ns} = e^{ih_n(s)}$$

and let $\nu_n^0$ be the left translation of $\nu_1^0$ by $z_n$, i.e. $\nu_n^0$ is the probability measure on $L(S^1)$ such that

$$\int_{L(S^1)} f(\sigma) d\nu_n^0(\sigma) = \int_{L(S^1)} f(z_n \sigma) d\nu_1^0(\sigma).$$

Also let $L_n(S^1)$ denote those $\sigma \in L(S^1)$ which are homotopic to $z_n$. 
**Remark 7.4.** The loops \(\{z_t\}_{t=0}^{T} \) are representatives from each of the homotopy classes of \(\mathcal{L}(S^1)\), i.e. \(\mathcal{L}(S^1)\) is the disjoint union of \(\{\mathcal{L}_n(S^1)\}_{n=-\infty}^{\infty}\). By the construction of \(\nu_t^0\) in Theorem 2.13, the measure \(\nu_t^0\) is concentrated on \(\mathcal{L}_0(S^1)\) and therefore \(\nu_t^0\) is concentrated on \(\mathcal{L}_n(S^1)\), i.e. \(\nu_t^0(\mathcal{L}_n(S^1)) = \delta_{mn}\).

**Proposition 7.5.** The relationship between pinned Wiener measure \(\mu_t^0\) and heat kernel measure \(\nu_t^0\) on \(\mathcal{L}(S^1)\) is

\[
\mu_t^0 = \frac{1}{q_t(1)} \sum_{n=-\infty}^{\infty} p_t(2\pi n) \nu_t^n.
\]

In particular

\[
(7.2) \quad \mu_t^0|_{\mathcal{L}_0(S^1)} = \frac{1}{q_t(1)} p_t(0) \nu_t^0 = \left( \sum_{n=-\infty}^{\infty} e^{-\frac{4t}{\pi^2}(2\pi n)^2} \right)^{-1} \nu_t^0.
\]

**Proof of Proposition 7.5.** To simplify notation, let \(B_s = \beta(t, s)\). Using Itô’s formula, one easily shows that the process \(\Sigma^0(t, s)\) in Theorem 2.13 and the process \(g\) in Eq. (2.5) of Definition 2.8 are given by \(\Sigma^0(t, s) = e^{\chi(t, s)}\) and \(g_s = e^{\delta(t, s)} = e^{B_s}\) respectively. Suppose that \(f : \mathcal{L}(S^1) \rightarrow \mathbb{R}\) is a cylinder function as in Definition 2.4, then for \(x \in (s_n, 1),\)

\[
\mu_t^0(f) = \mathbb{E} \left[ f(g) \frac{q_t(1-a)(g_\alpha)}{q_t(1)} \right] = \mathbb{E} \left[ f(e^{B_\alpha}) \frac{q_t(1-a)(e^{B_\alpha})}{q_t(1)} \right] = \frac{1}{q_t(1)} \mathbb{E} \left[ f(e^{B_\alpha}) \sum_{n=-\infty}^{\infty} p_t(1-a)(B_\alpha - 2\pi n) \right]
\]

\[
(7.3) = \frac{1}{q_t(1)} \sum_{n=-\infty}^{\infty} \mathbb{E} \left[ f(e^{B_\alpha}) p_t(1-a)(B_\alpha - 2\pi n) \right].
\]

Let \(h_n(s) := 2\pi n s, h_n^\alpha(s) = 2\pi n (s \land \alpha),\) and \(F(B) = f(e^{B_\alpha}),\) so that \(F\) is a bounded cylinder function \(W(\mathbb{R})\). By the Cameron-Martin theorem (making the translation \(B \rightarrow B + h_n^\alpha),\)

\[
\mathbb{E}[F(B)p_t(1-a)(B_\alpha - 2\pi n)] = \mathbb{E} \left[ F(B + h_n^\alpha)p_t(1-a)(B_\alpha - 2\pi n(1 - \alpha)) \right] \cdot \exp \left( -\frac{1}{t} \int_{0}^{\alpha} 2\pi n dB_s - \frac{1}{2t} \int_{0}^{\alpha} (2\pi n)^2 ds \right)
\]

\[
= \mathbb{E} \left[ F(B + h_n)p_t(1-a)(B_\alpha - 2\pi n(1 - \alpha)) \right] \cdot \exp \left( -\frac{2\pi n}{t} B_\alpha - \frac{1}{2t} \alpha(2\pi n)^2 \right).
\]

(7.4)

By direct computation,

\[
p_t(1-a)(x - y(1 - \alpha)) = p_t(1-a)(x) \cdot \exp \left( \frac{1}{t} xy - \frac{1}{2t}(1 - \alpha)y^2 \right)
\]

\[
(7.4)
\]
and thus taking \( x = B_\alpha \) and \( y = 2\pi n \),
\[
 p_{(1-\alpha)}(B_\alpha - 2\pi n(1-\alpha)) \cdot \exp \left( -\frac{2\pi n}{t} B_\alpha - \frac{1}{2t}(2\pi n)^2 \right)
\]
\[= p_{(1-\alpha)}(B_\alpha) \cdot \exp \left( -\frac{1}{2t}(2\pi n)^2 \right) .
\]
(7.5)

Combining Eqs. (7.4) and (7.5) shows that
\[
\mathbb{E}[F(B) \ p_{(1-\alpha)}(B_\alpha - 2\pi n)]
\]
\[= (2\pi t)^{-1/2} \exp \left( -\frac{1}{2t}(2\pi n)^2 \right) \mathbb{E} \left[ F(B + h_n) \frac{p_{(1-\alpha)}(B_\alpha)}{p_t(0)} \right]
\]
\[= p_t(2\pi n) \mathbb{E}[F(\chi(t, \cdot) + h_n)]
\]

wherein the second equality we have used Eq. (2.11) of Remark 2.12. Using this equation, with \( F(B) = f(e^{iB}) \), in Eq. (7.3) gives
\[
\mu_t^B(f) = \frac{1}{q_t(1)} \sum_{n=-\infty}^{\infty} p_t(2\pi n) \mathbb{E} \left[ f(e^{i(\chi(t, \cdot) + h_n)}) \right]
\]
\[= \frac{1}{q_t(1)} \sum_{n=-\infty}^{\infty} p_t(2\pi n) \mathbb{E} \left[ f(z_n e^{i\chi(t, \cdot)}) \right]
\]
\[= \frac{1}{q_t(1)} \sum_{n=-\infty}^{\infty} p_t(2\pi n) u_t^n(f).
\]

8. Appendix (Quadratic variations)

**Lemma 8.1.** As above, for \( A \in \mathcal{F} \) let \( \beta^A(t, s) = \langle \beta(t, s), A \rangle_t \) and \( \chi^A(t, s) = \langle \chi(t, s), A \rangle_t \). Let \( A, B \in \mathcal{F} \) and \( \sigma, \tau \in [0, 1] \), then
\[
\beta^A(dt, s)\beta^B(dt, \sigma) = \langle A, B \rangle_t G(s, \sigma) dt,
\]
\[
\chi^A(dt, s)\chi^B(dt, \sigma) = \langle A, B \rangle_t G_0(s, \sigma) dt,
\]
and for \( t, \tau \in [0, \infty) \),
\[
\beta^A(t, ds)\beta^B(t, ds) = \langle A, B \rangle_t G(t, \tau) ds.
\]

**Proof of Lemma 8.1.** Let \( \{\mathcal{G}_t\} \) be an abstract filtration (satisfying the “usual hypothesis”) and suppose that \( M_t \) and \( N_t \) are two continuous \( \mathcal{G}_t \) adapted processes such that \( (M_t - M_s, N_t - N_s) \) is independent of \( \mathcal{G}_s \) for all \( t > s \) and \( \mathbb{E}M_t = \mathbb{E}N_t = 0 \) for all \( t \geq 0 \). Then clearly \( M \) and \( N \) are \( \mathcal{G}_t \) martingales. We now also assert that
\[
M_t N_t - \mathbb{E}[M_t N_t] \text{ is a martingale}
\]
Assuming Eq. (8.1) for the moment, we may conclude the differential \( M_{dt} N_{dt} \) of the quadratic co-variation of \( M \) and \( N \) is given by
\[
M_{dt} N_{dt} = d_t \mathbb{E}[M_t N_t].
\]
The lemma then follows from repeated application of Eq. (8.2). For example, taking $M_t = \beta^A(t, s)$ and $N_t = \beta^B(t, \sigma)$, we learn that

$$
\beta^A(dt, s)\beta^B(dt, \sigma) = \frac{d_t}{d_s} G(s, \sigma) dt.
$$

To prove Eq. (8.1), let $t > s$, $M_t = M_{t-s}$, $N_t = N_{t-s}$, and $E_s = E(\cdot | \mathcal{G}_s)$. Then using the martingale properties of $M$ and $N$ and the independent increment assumption we find

$$
E_s [M_t N_t - M_s N_s] = E_s [(M_s + \Delta M)(N_s + \Delta N) - M_s N_s] = E_s [\Delta M \Delta N]
$$

$$
= E[(M_t - M_s)(N_t - N_s)] = E[(M_t - M_s)(N_t + N_s)]
$$

$$
= E[M_t N_t] - E[M_s N_s].
$$

Rearranging the terms of the result of this computation shows that

$$
E_s [M_t N_t - E[M_t N_t]] = M_s N_s - E[M_s N_s]
$$
as desired.

REFERENCES


