# DE BRUIJN IDENTITIES IN DIFFERENT MARKOVIAN CHANNELS 

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#### Abstract

De Bruijn's identity in information theory states that if $u$ is the solution of the heat equation, then the time derivative of the Shannon entropy for this solution is equal to the amount of Fisher information at $u$. In this article, we show how this identity changes if we replace the heat channel by the Fokker Planck, or passing from Fokker Planck to Ornstein-Uhlenbeck channels. Through these passages we investigate the different properties of these solutions. We exclusively dissect different properties of Ornstein-Uhlenbeck semigroup given by the Mehler formula expression.


## 1. Introduction

Let the probability triplet be $(\Omega, \mathscr{F}, \mu)$, where $\Omega$ is the sample space, $\mathscr{F}$ a $\sigma$ algebra and $\mu$ is a probability measure. Let $\varphi: \mathbb{R}^{n} \rightarrow \mathbb{R}_{+}$with $\int_{\mathbb{R}^{n}} \varphi(x) d \mu(x)=1$ be a density function of a random variable $X$. We extend the function $x \in(0, \infty) \mapsto$ $x \ln x$ by 0 at $x=0$ and assume that $\varphi \ln \varphi \in L^{1}\left(\mathbb{R}^{n}, d \mu\right)$. This function defines so called Shannon's entropy or Boltzmann $H$ function

$$
H(\varphi):=-\int_{\mathbb{R}^{n}} \varphi(x) \ln \varphi(x) d \mu(x)
$$

If we place ourselves in a dynamical system at time $t$, this entropy will be written as $H(\varphi(\cdot, t))$. In this context the term of de Bruijn identity was pointed by Stam [5], which was communicated to him by Prof. de Bruijn and indicate that Shannon's entropy decreases in time when $u$ runs through a Gaussian channel with rate equal to Fisher information.

In mathematical information theory, the Fisher information is a way of measuring the amount of information that an observable random variable $X$ carries about the distribution that models $\varphi$. For example by taking $X: \Omega \times[0, \infty) \mapsto \mathbb{R}$ a Markovian process and defining the density function $\varphi(\cdot, t)$ of the random variable valued in $\mathbb{R}_{+}$can be considered as a probability distribution depending on $t \in \mathbb{R}_{+}$. Formally, the Fisher information is the variance of the score, which is the gradient of the loglikelihood function which is logarithm of $\varphi(\cdot, t)$. This is the fundamental concept

[^0]in information theory as it is indicated in the seminal book of Cover and Thomas [2].

The Fisher information is defined by the following quantity in $[0, \infty]$

$$
\begin{aligned}
\mathcal{I}(\varphi) & =\int_{\mathbb{R}^{n}}\left(\frac{\partial}{\partial x} \ln \varphi(x, t)\right)^{2} \varphi(x, t) d \mu(x) \\
& =\int_{\mathbb{R}^{n}}\left(\frac{\partial}{\partial x} \varphi(x, t)\right)^{2} \varphi(x, t)^{-1} d \mu(x)
\end{aligned}
$$

The most well-known example in a dynamical system is the unitary heat equation

$$
\begin{equation*}
\frac{\partial u}{\partial t}=\Delta u, \quad t>0, x \in \mathbb{R}^{n} \tag{1.1}
\end{equation*}
$$

with an initial condition $u_{0}(\cdot)=u(\cdot, 0) \in L_{+}^{1}\left(\mathbb{R}^{n}, d \mu\right)$ where $d \mu=d x$ is the standard Lebesgue measure with

$$
\begin{equation*}
\int_{\mathbb{R}^{n}} u_{0}(x) d x=1 \tag{1.2}
\end{equation*}
$$

As a straightforward consequence of the explicit expression of the solution in term of the Green function $G(t, x, y):=(4 \pi t)^{-n / 2} \exp \left(-\frac{|x-y|^{2}}{4 t}\right)$,

$$
\begin{equation*}
u(x, t)=\int_{\mathbb{R}^{n}} u_{0}(y) G(t, x, y) d y \tag{1.3}
\end{equation*}
$$

it follows that $u(x, t)$ is positive for any $(x, t) \in \mathbb{R}^{n} \times \mathbb{R}_{+}$. The mass conservation implies also that

$$
\begin{equation*}
\int_{\mathbb{R}^{n}} u(x, t) d x=\int_{\mathbb{R}^{n}} u_{0}(x) d x=1 \tag{1.4}
\end{equation*}
$$

In this case we say that $u$ runs through the Gaussian or heat channel.
The noticeable connection between Fisher information and Shannon's entropy is the so-called De Bruijn relation [2] (see also [1, 4]). That is, if $u$ runs through the Gaussian channel, then

$$
\begin{equation*}
\frac{d}{d t} H(u)=\mathcal{I}(u) \tag{1.5}
\end{equation*}
$$

(For completeness, the proof is provided in the Appendix).
Before trying to deduce the De Bruijn relation for Fokker-Planck equation

$$
\frac{\partial}{\partial t} v(x, t)=\frac{\partial^{2}}{\partial x^{2}}(v(x, t))-\frac{\partial}{\partial x}(x v(x, t))
$$

in the next section we show how one can derive the Fokker-Planck equation from the heat equation and in the section 3 we establish the De Bruijn relation for Fokker-Planck equation.

In section 4 we show how the Ornstein-Uhlenbeck equation $\frac{\partial w}{\partial t}=\frac{\partial^{2}}{\partial y^{2}} w(x, t)-$ $x \frac{\partial}{\partial x} w(x, t)$ can be deduce from Fokker-Planck equation. For this equation the mass conservation takes place in $L^{1}\left(\mathbb{R}^{n}, d \mu\right)$, where $d \mu$ is the Gaussian measure. The section 5 is devoted to Ornstein-Uhlenbeck semigroup in which we prove the hypercontractivity of this semigroup which deduces the Chapman-Kolmogorov relation for its kernel. In the section 6 we recover the De Bruijn relation for this channel.

Finally in section 7 we prove the De Bruijn identity for relative Fisher information and Kullback-Leibler divergence which is already discussed in [7].

## 2. Relationship between Fokker-Planck and heat equation

In general the Fokker-Planck equation in one dimensional space reads as

$$
\begin{equation*}
\frac{\partial}{\partial t} v(y, \tau)=\frac{\partial^{2}}{\partial y^{2}}\left(g^{2}(y, t) v(y, \tau)\right)-\frac{\partial}{\partial y}(f(y) v(y, \tau)) \tag{2.1}
\end{equation*}
$$

where $f(y, t)$ and $g(y, \tau)$ can be arbitrary positive functions define on $\mathbb{R}_{y} \times \mathbb{R}_{+}$. In this section we take $g=1$ and $f(y)=-y$ and the following theorem gives an explicit expression of the solution of (2.1).

Theorem 2.1. If $u$ is the solution of the heat equation (1.1) with initial condition $u_{0}$ satisfying 1.2), then $v(y, \tau)=e^{\tau} u\left(e^{\tau} y,\left(e^{2 \tau}-1\right) / 2\right)$, that is

$$
\begin{equation*}
v(y, \tau)=\frac{e^{\tau}}{\sqrt{2 \pi\left(e^{2 \tau}-1\right)}} \int_{\mathbb{R}} v_{0}(\xi) e^{-\frac{\left(e^{\tau} y-\xi\right)^{2}}{2\left(e^{2 \tau}-1\right)}} d \xi \tag{2.2}
\end{equation*}
$$

satisfies the Fokker-Planck equation

$$
\begin{equation*}
\frac{\partial}{\partial \tau} v(y, \tau)=\frac{\partial^{2}}{\partial y^{2}} v(y, \tau)+v(y, \tau)+y \frac{\partial}{\partial y} v(y, \tau) \tag{2.3}
\end{equation*}
$$

Proof. First we remark that for $t=\left(e^{2 r}-1\right) / 2=0$, we have $e^{2 \tau}-1=0$, so $\tau$ should be equal zero. Hence

$$
\begin{equation*}
u_{0}=v(y, 0)=v_{0} . \tag{2.4}
\end{equation*}
$$

If we replace $u(x, t)$ by its explicit expression $(1.3)$ and we find 2.2$)$. Now we have to verify that this function bears out (2.3). Indeed, let us denote

$$
\begin{gathered}
A(\tau)=\frac{e^{\tau}}{\sqrt{2 \pi\left(e^{2 \tau}-1\right)}}, \quad B:=B(\tau, y, \xi)=\frac{\left(e^{\tau} y-\xi\right)^{2}}{2\left(e^{2 \tau}-1\right)} \\
I(\tau, y)=\int_{\mathbb{R}} v_{0}(\xi) e^{-B(\tau, y, \xi)} d \xi
\end{gathered}
$$

such that 2.2 can be expressed as

$$
v(y, \tau)=A(\tau) I(\tau, y)
$$

We remark that

$$
\begin{gathered}
\frac{\partial}{\partial \tau} A(\tau)=\underbrace{\frac{e^{\tau}}{\sqrt{2 \pi\left(e^{2 \tau}-1\right)}}-\underbrace{\frac{e^{3 \tau}}{\sqrt{2 \pi\left(e^{2 \tau}-1\right)^{3}}}}_{=A_{2}(\tau)}}_{=A_{1}(\tau)} \begin{array}{c}
\frac{\partial}{\partial \tau} B(\tau, y, \xi)=\underbrace{\frac{y e^{\tau}\left(e^{\tau} y-\xi\right)}{e^{2 \tau}-1}}_{=B_{1}(\tau, y, \xi)}-\underbrace{\frac{e^{2 \tau}\left(e^{\tau} y-\xi\right)^{2}}{\left(e^{2 \tau}-1\right)^{2}}}_{=B_{2}(\tau, y, \xi)} \\
\frac{\partial}{\partial y} B(\tau, y, \xi)=\underbrace{\frac{e^{\tau}\left(e^{\tau} y-\xi\right)}{e^{2 \tau}-1}}_{=B_{3}(\tau, y, \xi)} \\
\frac{\partial}{\partial y} B_{3}(\tau, y, \xi)=\frac{e^{2 \tau}}{e^{2 \tau}-1}
\end{array},
\end{gathered}
$$

Hence,

$$
\frac{\partial}{\partial y} v(y, \tau)=-A(\tau) \int_{\mathbb{R}} v_{0}(\xi) B_{3}(\tau, y, \xi) e^{-B(\tau, y, \xi)} d \xi
$$

and

$$
\begin{aligned}
\frac{\partial^{2}}{\partial y^{2}} v(y, \tau)= & A(\tau)\left(-\int_{\mathbb{R}} v_{0}(\xi) \frac{e^{2 \tau}}{e^{2 \tau}-1} e^{-B(\tau, y, \xi)} d \xi\right. \\
& \left.+\int_{\mathbb{R}} v_{0}(\xi)\left(B_{3}(\tau, y, \xi)\right)^{2} e^{-B(\tau, y, \xi)} d \xi\right) \\
= & -A_{2}(\tau) I(\tau, y)+A(\tau) \int_{\mathbb{R}} u_{0}(\xi)\left(B_{3}(\tau, y, \xi)\right)^{2} e^{-B(\tau, y, \xi)} d \xi \\
= & -A_{2}(\tau) I(\tau, y)+A(\tau) \int_{\mathbb{R}} u_{0}(\xi) B_{2}(\tau, y, \xi) e^{-B(\tau, y, \xi)} d \xi
\end{aligned}
$$

Consequently,

$$
\begin{aligned}
& \frac{\partial}{\partial \tau} v(y, \tau) \\
& =\underbrace{A_{1}(\tau) I(\tau, y)}_{=v(y, \tau)}-A_{2}(\tau) I(\tau, y) \\
& \quad \underbrace{-A(\tau)\left(\int_{\mathbb{R}} u_{0}(\xi) B_{1}(\tau, y, \xi) e^{-B(\tau, y, \xi)} d \xi\right.}_{=y \frac{\partial v(y, \tau)}{\partial y}}-\int_{\mathbb{R}} v_{0}(\xi) B_{2}(\tau, y, \xi) e^{-B(\tau, y, \xi)} d \xi) \\
& =v(y, \tau)+y \frac{\partial v(y, \tau)}{\partial y}+\frac{\partial^{2}}{\partial y^{2}} v(y, \tau) .
\end{aligned}
$$

## 3. De Bruijn identity in Fokker-Planck channels

In this section we will use the above Theorem for obtaining an identity similar to (1.5).

Theorem 3.1. Assume that $t \geq 0$, the Fisher information is defined by the positive quantity

$$
\mathcal{I}(v(\cdot, t))=\int_{\mathbb{R}}\left(\frac{\partial}{\partial y} \ln v(y, t)\right)^{2} v(y, t) d y
$$

and Shannon's entropy is

$$
\begin{equation*}
H(v(\cdot, t))=-\int_{\mathbb{R}} v(y, t) \ln v(y, t) d y \tag{3.1}
\end{equation*}
$$

Then $v:=v(y, \tau)$ the solution of

$$
\begin{align*}
\frac{d}{d \tau} v(y, \tau) & =\frac{\partial^{2}}{\partial y^{2}} v(y, \tau)+v(y, \tau)+y \frac{\partial}{\partial y} v(y, \tau)  \tag{3.2}\\
v(y, 0) & =v_{0}(y), \quad \text { with } \int_{\mathbb{R}} v_{0}(y) d y=1
\end{align*}
$$

will satisfy the modified De Bruijn identity in Fokker-Planck channels

$$
\begin{equation*}
\frac{d}{d \tau} H(v)=\mathcal{I}(v)-1 \tag{3.3}
\end{equation*}
$$

Proof. According 2.2,

$$
\begin{equation*}
\lim _{|y| \rightarrow \infty}|y v|=0 \tag{3.4}
\end{equation*}
$$

and

$$
\begin{aligned}
\frac{d}{d \tau} H(v(\cdot, \tau))= & -\int_{\mathbb{R}}\left(\frac{\partial^{2}}{\partial y^{2}} v(y, \tau)+\frac{\partial}{\partial y}(y v(y, \tau))\right) \ln v(y, \tau) d y \\
& -\int_{\mathbb{R}}\left(\frac{\partial^{2}}{\partial y^{2}} v(y, \tau)+\frac{\partial}{\partial y}(y v(y, \tau))\right) d y \\
= & \int_{\mathbb{R}}\left(\left(\frac{\partial v}{\partial y}\right)^{2} / v+y \frac{\partial v}{\partial y}(y, \tau)\right) d y \\
= & \mathcal{I}(v)-\int_{\mathbb{R}} v(y, \tau) d y=\mathcal{I}(v)-1
\end{aligned}
$$

according to (3.4), Fokker-Planck equation has the mass conservation property and for $t=0$ the mass is equal 1 , that is

$$
\begin{equation*}
\int_{\mathbb{R}} v(y, \tau) d y=1 \quad \text { for all } \tau \in \mathbb{R}_{+} \tag{3.5}
\end{equation*}
$$

## 4. Relationship between Fokker-Planck and Ornstein-Uhlenbeck EQUATION

Let $v_{\infty}(y):=\frac{1}{\sqrt{2 \pi}} e^{-y^{2} / 2}$ be the unique stationary solution of Fokker-Planck equation 2.3 with $\int_{\mathbb{R}} v_{\infty}(y) d y=1$, and denote by $d \mu=v_{\infty} d x$ the Gaussian measure. Now, if we transform the expression (2.2) to the form

$$
\begin{equation*}
v(y, \tau)=\frac{1}{\sqrt{2 \pi\left(1-e^{-2 \tau}\right)}} \int_{\mathbb{R}} v_{0}(\xi) e^{-\frac{\left(y-\xi / e^{\tau}\right)^{2}}{2\left(1-e^{-2 \tau}\right)}} d \xi \tag{4.1}
\end{equation*}
$$

we notice that $v(y, \tau) \rightarrow v_{\infty}(y)$ as $\tau \rightarrow \infty$.
Theorem 4.1. Assume that $v$ is the solution of Fokker-Planck equation (2.3). Then $w=v / v_{\infty}$ satisfies the Ornstein-Uhlenbeck equation

$$
\begin{equation*}
\frac{\partial w}{\partial \tau}=\frac{\partial^{2}}{\partial y^{2}} w(y, \tau)-y \frac{\partial}{\partial y} w(y, \tau) \tag{4.2}
\end{equation*}
$$

with initial data $w_{0}(y)=u_{0}(y) / v_{\infty}, u_{0} \in L_{+}^{1}\left(\mathbb{R}^{n}, d y\right)$ and $\int_{\mathbb{R}^{n}} u_{0}(y) d y=1$.
Proof. First note that by 2.4, $w_{0}=w(y, 0)=v_{0}(y) / v_{\infty}(y)=u_{0}(y) / v_{\infty}(y)$. Consequently, by 3.5 we have $\int_{\mathbb{R}} w(y, \tau) d \mu=1$.

Now, knowing that $v$ is the solution of (3.1) we can write

$$
\begin{equation*}
\frac{\partial w}{\partial \tau}=\left(\frac{\partial^{2} v}{\partial y^{2}}+v+y \frac{\partial v}{\partial y}\right) / v_{\infty} \tag{4.3}
\end{equation*}
$$

On the other hand, by $\frac{\partial v_{\infty}}{\partial y}=-y v_{\infty}$, we have

$$
\begin{gathered}
\frac{\partial w}{\partial y}=v_{\infty}^{-1}\left(\frac{\partial v}{\partial y}+y v\right) \\
\frac{\partial^{2} w}{\partial y^{2}}=v_{\infty}^{-1}\left(\frac{\partial^{2} v}{\partial y^{2}}+2 y \frac{\partial v}{\partial y}+v+y^{2} v\right)
\end{gathered}
$$

By insert these expressions in 4.3 we obtain 4.2 .

## 5. Mehler formula and Ornstein-Uhlenbeck semigroup

On $\mathbb{R}^{n}$, let $\mu_{n}$ be the canonical Gaussian measure with density $(2 \pi)^{-n / 2} e^{\left(-|x|^{2} / 2\right)}$ with respect to the Lebesgue measure $d x$. With this measure we consider the Banach space $L^{p}\left(\mathbb{R}^{n}, d \mu_{n}\right), 0 \leq p<\infty$ with the norm $\|f\|_{p}=\left(\int_{\mathbb{R}^{n}}|f|^{p} d \mu_{n}\right)^{1 / p}$ on which we can define the Ornstein-Uhlenbeck semigroup $P_{t}$ by mean of Mehler formula

$$
\begin{equation*}
P_{t} f(x):=\int_{\mathbb{R}^{n}} f\left(e^{-t} x+\sqrt{1-e^{-2 t}} y\right) d \mu_{n}(y), \quad \text { for } f \in L^{p}\left(\mathbb{R}^{n}, d \mu_{n}\right) \tag{5.1}
\end{equation*}
$$

By taking $\alpha_{t}=e^{-t}, \beta_{t}=\sqrt{1-e^{-2 t}}$ and making a change of variable $z=\alpha_{t} x+\beta_{t} y$ in this formula we obtain

$$
\begin{equation*}
P_{t} f(x)=\left(2 \pi \beta_{t}^{2}\right)^{-n / 2} \int_{\mathbb{R}^{n}} f(z) \exp \left(-\left|z-\alpha_{t} x\right|^{2} / 2\left(\beta_{t}^{2}\right)\right) d z \tag{5.2}
\end{equation*}
$$

This equality implies that the Gaussian measure $d \mu_{n}$ is invariant for $P_{t}$, that is

$$
\begin{equation*}
\int_{\mathbb{R}^{n}} P_{t} f(x) d \mu_{n}(x)=\int_{\mathbb{R}^{n}} f(x) d \mu_{n}(x) \quad \text { for all } f \in L^{p}\left(\mathbb{R}^{n}, d \mu_{n}\right) \tag{5.3}
\end{equation*}
$$

To show this we need the following lemma.
Lemma 5.1. For $c_{1}, c_{2} \geq 0, c_{1}+c_{2} \neq 0$ and $a, b \in \mathbb{R}^{n}$, we have

$$
\begin{equation*}
\int_{\mathbb{R}^{n}} e^{-c_{1}|a-z|^{2}-c_{2}|z-b|^{2}} d z=\left(\frac{\pi}{c_{1}+c_{2}}\right)^{n / 2} \exp \left(-\frac{c_{1} c_{2}}{c_{1}+c_{2}}|a-b|^{2}\right) \tag{5.4}
\end{equation*}
$$

Proof. This follows from the unity of the Gaussian measure that for any $p \in \mathbb{R}^{n}$ and $\alpha>0$,

$$
\int_{\mathbb{R}^{n}} \exp \left(-\alpha|x-p|^{2}\right) d x=\left(\frac{\pi}{\alpha}\right)^{n / 2}
$$

which implies

$$
\int_{\mathbb{R}^{n}} \exp \left(-\alpha|x|^{2}+2\langle\alpha p, x\rangle\right) d y=\left(\frac{\pi}{\alpha}\right)^{n / 2} \exp \left(\alpha|p|^{2}\right)
$$

Let $\alpha=c_{1}+c_{2}$ and $p=\left(c_{1} a+c_{2} b\right) / \alpha$, then

$$
\begin{aligned}
& \int_{\mathbb{R}^{n}} \exp \left(-\left(c_{1}+c_{2}\right)|x|^{2}+2\left\langle\left(c_{1} a+c_{2} b\right), x\right\rangle\right) d x \\
& =\left(\frac{\pi}{c_{1}+c_{2}}\right)^{n / 2} \exp \left(\frac{\left|\left(c_{1} a+c_{2} b\right)\right|^{2}}{c_{1}+c_{2}}\right)
\end{aligned}
$$

Now since

$$
\begin{aligned}
& \int_{\mathbb{R}^{n}} \exp \left(-c_{1}|a-x|^{2}-c_{2}|x-b|^{2}\right) d x \\
& =\left(\frac{\pi}{c_{1}+c_{2}}\right)^{n / 2} \exp \left(\frac{\left|\left(c_{1} a+c_{2} b\right)\right|^{2}}{c_{1}+c_{2}}-c_{1}|a|^{2}-c_{2}|b|^{2}\right) \\
& =\left(\frac{\pi}{c_{1}+c_{2}}\right)^{n / 2} \exp \left(\frac{-c_{1} c_{2}|a-b|^{2}}{c_{1}+c_{2}}\right),
\end{aligned}
$$

we obtain (5.4).

For proving (5.3 we write
$\int_{\mathbb{R}^{n}} P_{t} f(x) d \mu_{n}(x)=\left(4 \pi^{2} \beta_{t}^{2}\right)^{-n / 2} \int_{\mathbb{R}^{n}} \int_{\mathbb{R}^{n}} f(z) \exp \left(-\frac{\left|z-\alpha_{t} x\right|^{2}+\left|\beta_{t} x\right|^{2}}{2 \beta_{t}^{2}}\right) d x d z$.
Taking $c_{1}=\frac{\alpha_{t}^{2}}{2 \beta_{t}^{2}}, c_{2}=\frac{1}{2}, a=\alpha_{t}^{-1}$ and $b=0$ in Lemma 5.1. since $\alpha_{t}^{2}+\beta_{t}^{2}=1$ we obtain

$$
\int_{\mathbb{R}^{n}} P_{t} f(x) d \mu_{n}(x)=(2 \pi)^{-n / 2} \int_{\mathbb{R}^{n}} f(z) \exp \left(-\frac{|z|^{2}}{2}\right) d z
$$

which is (5.3).
Theorem 5.2. On $X_{p}:=L^{p}\left(\mathbb{R}^{n}, d \mu_{n}\right)$ the operator $T_{t}$ defines a hypercontractive semigroup; that is,
(i) $\left\|T_{t}\right\|_{p} \leq\|f\|_{q}$, for all $p \geq q>1$ such that $p-1 \leq e^{2 t}(q-1)$;
(ii) $\lim _{t \rightarrow 0}\left\|T_{t} f-f\right\|_{p}=0$, for all $f \in X_{p}$;
(iii) $T_{t} T_{s}=T_{t+s}$, for all $(t, s) \in \mathbb{R}_{+}^{2}$.

Proof. (i) Since the constant function 1is in $X_{p}$, and the Mehler formula $P_{t} 1=1$, the equality (5.3) implies that $P_{t}$ is doubly Markovian in the sense of Nelson (see [3]). In the same paper (Theorem 2), Nelson shows that an operator which is doubly Markovian is hypercontractive in the sense of item (i).
(ii) Since $\alpha_{t}^{2}+\beta_{t}^{2}=1$, the vectors $\left(\alpha_{t}, \beta_{t}\right)$ and $(1,0)$ are both on the unit circle and $\left(\alpha_{t}, \beta_{t}\right) \rightarrow(1,0)$ as $t \rightarrow 0$. For any continuous bounded function $f$ (taking e.g. $f \in \mathscr{S})$,

$$
f\left(\alpha_{t} x+\beta_{t} y\right)-f(x)=f\left(\left(\alpha_{t}, \beta_{t}\right)\binom{x}{y}\right)-f\left((1,0)\binom{x}{y}\right) \rightarrow 0
$$

as $t \rightarrow 0$. Furthermore, if $M=\sup _{x \in \mathbb{R}}|f(x)|$, then $\left|f\left(\alpha_{t} x+\beta_{t} y\right)-f(x)\right| \leq 2 M$ in $L^{1}\left(\mathbb{R}^{n}, d \mu_{n}\right)$. Thus according to Lebesgue's dominated converence theorem

$$
\int_{\mathbb{R}^{n}}\left|f\left(\alpha_{t} x+\beta_{t} y\right)-f(x)\right| d \mu_{n}(y) \rightarrow 0
$$

Hence

$$
\begin{aligned}
\left\|T_{t} f-f\right\|_{p}^{p} & =\int_{\mathbb{R}^{n}}\left|P_{t} f(x)-f(x)\right|^{p} d \mu_{n}(x) \\
& =\int_{\mathbb{R}^{n}}\left|\int_{\mathbb{R}^{n}}\left[f\left(\alpha_{t} x+\beta_{t} y\right)-f(x)\right] d \mu_{n}(y)\right|^{p} d \mu_{n}(x) \rightarrow 0
\end{aligned}
$$

Since the Schwartz space $\mathscr{S}$ being dense in $X_{p}$, this implies item (ii).
(iii) Taking the expression of the Ornstein-Uhlenbeck semigroup $P_{t}$ (5.2),

$$
\begin{aligned}
P_{t} P_{s} f(x) & =\left(2 \pi \beta_{t}^{2}\right)^{-n / 2} \int_{\mathbb{R}^{n}} P_{s} f(z) \exp \left(-\left|z-\alpha_{t} x\right|^{2} / 2\left(\beta_{t}^{2}\right) d z\right. \\
& =\left(2 \pi \beta_{t} \beta_{s}\right)^{-n} \int_{\mathbb{R}^{n}} f(y) \underbrace{\int_{\mathbb{R}^{n}} \exp \left(-\frac{\left|y-\alpha_{s} z\right|^{2}}{2 \beta_{s}^{2}}-\frac{\left|z-\alpha_{t} x\right|^{2}}{2 \beta_{t}^{2}}\right) d z}_{=A} d y
\end{aligned}
$$

To simplify the expression $A$ we will use the Lemma 5.1. Let

$$
A=\int_{\mathbb{R}^{n}} \exp \left(-\frac{\alpha_{s}^{2}}{2 \beta_{s}^{2}}\left|\alpha_{s}^{-1} y-z\right|^{2}-\frac{1}{2 \beta_{t}^{2}}\left|z-\alpha_{t} x\right|^{2}\right) d z
$$

Comparing this with 5.4, we obtain

$$
c_{1}=\frac{\alpha_{s}^{2}}{2 \beta_{s}^{2}}, \quad c_{2}=\frac{1}{2 \beta_{t}^{2}}, \quad a=\alpha_{s}^{-1} y, \quad b=\alpha_{t} x
$$

Hence,

$$
\begin{align*}
A & =\left(\frac{\pi}{\frac{\alpha_{s}^{2}}{2 \beta_{s}^{2}}+\frac{1}{2 \beta_{t}^{2}}}\right)^{n / 2} \exp \left(-\frac{\frac{\alpha_{s}^{2}}{4 \beta_{s}^{2} \beta_{t}^{2}}}{\frac{\alpha_{s}^{2}}{2 \beta_{s}^{2}}+\frac{1}{2 \beta_{t}^{2}}}\left|\alpha_{s}^{-1} y-\alpha_{t} x\right|^{2}\right)  \tag{5.5}\\
& =\left(\frac{2 \pi\left(1-e^{-2 t}\right)\left(1-e^{-2 s}\right)}{1-e^{-2(t+s)}}\right)^{n / 2} \exp \left(-\frac{\left|y-e^{-(t+s)} x\right|^{2}}{2\left(1-e^{-2(t+s)}\right)}\right)
\end{align*}
$$

Replacing the expression of $A$ in 5.5 we find that

$$
P_{t} P_{s} f(x)=\left(2 \pi\left(1-e^{-2(t+s)}\right)\right)^{-\frac{n}{2}} \int_{\mathbb{R}^{n}} f(y) \exp \left(-\frac{\left|y-e^{-(t+s)} x\right|^{2}}{2\left(1-e^{-2(t+s)}\right)}\right) d y=P_{t+s} f(x)
$$

Remark 5.3. From (iii) of the above Theorem one can deduce the ChapmanKolmogorov formula

$$
\begin{equation*}
\int_{\mathbb{R}^{n}} K(x, y, t) K(y, z, s) d y=K(x, z, t+s) \quad \text { for all } x \in \mathbb{R}^{n}, t, s>0 \tag{5.6}
\end{equation*}
$$

where $K(x, y, t)$ is the heat kernel of Ornstein-Uhlenbeck semigroup, that is

$$
P_{t} f(x):=\int_{\mathbb{R}^{n}} K(x, y, t) f(y) d y \quad \text { for all } f \in L^{p}\left(\mathbb{R}^{n}, d \mu_{n}\right)
$$

From 5.2 it follows that

$$
K(x, y, t)=\left(2 \pi\left(1-t^{-2 t}\right)\right)^{-n / 2} \exp \left(-\frac{\left|e^{-t} x-y\right|^{2}}{2\left(1-t^{-2 t}\right)}\right)
$$

Hence

$$
\begin{aligned}
P_{t} P_{s} f(x) & =\int_{\mathbb{R}^{n}} K(x, y, t) P_{s} f(y) d y \\
& =\iint_{\mathbb{R}^{2 n}} K(x, y, t) K(y, z, s) f(z) d z d y \\
& =P_{t+s} f(x)=\int_{\mathbb{R}^{n}} K(x, z, t+s) f(z) d z
\end{aligned}
$$

Since this identity holds for all $f \in L^{p}\left(\mathbb{R}^{n}, d \mu_{n}\right)$, we deduce formula (5.6) for $\mu_{n^{-}}$ a.e. $z \in \mathbb{R}^{n}$. The equality holds on $\mathbb{R}^{n}$ by continuity of the left and right hand side with respect to $z$.

## 6. De Bruijn identity in Ornstein-Uhlenbeck channels

In this section we work in $L^{1}(\mathbb{R}, \mu)$ which is a Lebesgue space with the Gaussian measure $\mu:=\mu_{1}$. In this space $\int_{\mathbb{R}} w(\cdot, \tau) d \mu=1$. If we define the entropy by

$$
\begin{equation*}
H_{\mu}(w(\cdot, \tau)):=-\int_{\mathbb{R}} w(y, \tau) \ln w(y, \tau) d \mu(y) \tag{6.1}
\end{equation*}
$$

and the Fisher information by

$$
\begin{equation*}
\mathcal{I}_{\mu}(w(\cdot, \tau)):=\int_{\mathbb{R}} w(y, \tau)\left(\frac{\partial}{\partial y} \ln w(y, \tau)\right)^{2} d \mu(y) \tag{6.2}
\end{equation*}
$$

then the De Bruijn identity in Ornstein-Uhlenbeck channels reads as follows.
Theorem 6.1. Assume that $t \geq 0$, then

$$
\begin{equation*}
\frac{d}{d \tau} H_{\mu}(w(\cdot, \tau))=\mathcal{I}_{\mu}(w(\cdot, \tau)) \tag{6.3}
\end{equation*}
$$

where $w$ is the solution of Ornstein-Uhlenbeck equation 4.2.
Proof. For the proof we will use

$$
\begin{equation*}
\frac{d}{d y} \mu(y)=\frac{-y}{\sqrt{2 \pi}} e^{-y^{2} / 2}=-y v_{\infty} \tag{6.4}
\end{equation*}
$$

The derivative of (6.1) with respect to $\tau$ reads

$$
\begin{aligned}
& \frac{d}{d \tau} H_{\mu}(w(y, \tau)) \\
& =-\int_{\mathbb{R}}\left(\frac{d}{d \tau} w(y, \tau) \ln w(y, \tau)\right) d \mu \\
& =-\int_{\mathbb{R}}\left(\frac{\partial^{2}}{\partial y^{2}} w(y, \tau)-y \frac{\partial}{\partial y} w(y, \tau)\right) \ln w(y, \tau) d \mu \underbrace{=}_{=0 \text { by } \sqrt[3.5]{-\frac{d}{d \tau} \int_{\mathbb{R}}} w(y, \tau) d \mu} \\
& \quad \int_{\mathbb{R}} w(y, \tau)^{-1}\left(\frac{\partial}{\partial y} w(y, \tau)\right)^{2} d \mu(y)-\int_{\mathbb{R}} \frac{\partial}{\partial y} w(y, \tau)(\ln w(y, \tau)) y v_{\infty} d y \\
& \quad+\int_{\mathbb{R}} y \frac{\partial}{\partial y} w(y, \tau)(\ln w(y, \tau)) d \mu=\mathcal{I}_{\mu}(w(y, \tau))
\end{aligned}
$$

## 7. De Bruijn identity for relative Fisher information and Kullback-Leibler divergence

Let $\varphi$ and $\psi$, be two distribution functions for two random variables $X$ and $Y$. The relative Fisher information with respect to $a$ is defined by

$$
\mathcal{I}_{a}(\varphi \| \psi):=\int_{\mathbb{R}} \varphi(x)\left(\frac{\partial}{\partial x} \ln \frac{\varphi(x)}{\psi(x)}\right)^{2} a(x) d x
$$

We define the Kullback-Leibler divergence which can be interpreted as the relative entropy between $\varphi$ and $\psi$ by

$$
D_{K L}(\varphi \| \psi):=\int_{\mathbb{R}} \varphi(x) \ln \frac{\varphi(x)}{\psi(x)} d x
$$

(see [6]).
The following result establishes that the relative entropy between any two solutions of 2.3 is always decreasing, with a rate given by the relative Fisher information:

Theorem 7.1. Assume that $\varphi(x, t)$ and $\psi(x, t)$ two distinct solutions of the FokkerPlanck equation in its general form:

$$
\begin{equation*}
\frac{\partial \phi}{\partial t}(x, t)=\frac{\partial^{2}}{\partial x^{2}}(a(x, t) \phi(x, t))-\frac{\partial}{\partial x}(b(x, t) \phi(x, t)) \tag{7.1}
\end{equation*}
$$

Then

$$
\begin{equation*}
\frac{d}{d t} D_{K L}(\varphi \| \psi)=-\mathcal{I}_{b}(\varphi \| \psi) \tag{7.2}
\end{equation*}
$$

Proof. Let $(x, t) \in \mathbb{R} \times \mathbb{R}_{+}$and assume $\varphi(\cdot, t), \psi(\cdot, t) \in H^{2}(\mathbb{R})$. By differentiating under the integral sign and using the chain rule, we have (For simplicity we write $\varphi(x)$ instead of $\varphi(x, t))$

$$
\begin{align*}
& \frac{d}{d t} D_{K L}(\varphi \| \psi) \\
& =\frac{d}{d t} \int_{\mathbb{R}} \varphi(x) \ln \frac{\varphi(x)}{\psi(x)} d x \\
& =\int_{\mathbb{R}} \frac{\partial}{\partial t} \varphi(x) \ln \frac{\varphi(x)}{\psi(x)} d x+\int_{\mathbb{R}} \varphi(x) \frac{\partial}{\partial t} \ln \varphi(x) d x-\int_{\mathbb{R}} \varphi(x) \frac{\partial}{\partial t} \ln \psi(x) d x  \tag{7.3}\\
& =\int_{\mathbb{R}} \frac{\partial}{\partial t} \varphi(x) \ln \frac{\varphi(x)}{\psi(x)} d x+0-\int_{\mathbb{R}} \frac{\varphi(x)}{\psi(x)} \frac{\partial}{\partial t} \psi(x) d x
\end{align*}
$$

By replacing $\frac{\partial}{\partial t} \varphi(x)$ in the Fokker-Planck equation 7.1 and using integration by parts we can write the first integral in 7.3 as

$$
\begin{align*}
& \int_{\mathbb{R}} \frac{\partial}{\partial t} \varphi(x) \ln \frac{\varphi(x)}{\psi(x)} d x \\
& =\int_{\mathbb{R}}\left(\frac{\partial^{2}}{\partial x^{2}} a(x) \varphi(x)-\frac{\partial}{\partial x} b(x) \varphi(x)\right) \ln \frac{\varphi(x)}{\psi(x)} d x  \tag{7.4}\\
& =\int_{\mathbb{R}}\left(a(x) \varphi(x) \frac{\partial^{2}}{\partial x^{2}} \ln \frac{\varphi(x)}{\psi(x)}+b(x) \varphi(x) \frac{\partial}{\partial x} \ln \frac{\varphi(x)}{\psi(x)}\right) d x .
\end{align*}
$$

Since

$$
\frac{\partial}{\partial x} \ln \frac{\varphi(x)}{\psi(x)}=\left(\frac{\psi(x)}{\varphi(x)}\right)\left(\frac{\partial}{\partial x} \frac{\varphi(x)}{\psi(x)}\right)
$$

and

$$
\frac{\partial^{2}}{\partial x^{2}} \ln \frac{\varphi(x)}{\psi(x)}=\left(\frac{\psi(x)}{\varphi(x)}\right)\left(\frac{\partial^{2}}{\partial x^{2}} \frac{\varphi(x)}{\psi(x)}\right)-\left(\frac{\partial}{\partial x} \ln \frac{\varphi(x)}{\psi(x)}\right)^{2}
$$

by replacing these relations in (7.4) and using integration by parts and the FokkerPlanck equation (7.1) for $\psi(x, t)$ we find that

$$
\begin{aligned}
& \int_{\mathbb{R}} \frac{\partial}{\partial t} \varphi(x) \ln \frac{\varphi(x)}{\psi(x)} d x \\
& =\int_{\mathbb{R}} a(x) \varphi(x)\left(\frac{\psi(x)}{\varphi(x)} \frac{\partial^{2}}{\partial x^{2}} \frac{\varphi(x)}{\psi(x)}-\left(\frac{\partial}{\partial x} \ln \frac{\varphi(x)}{\psi(x)}\right)^{2}\right) d x+\int_{\mathbb{R}} b(x) \psi(x) \frac{\partial}{\partial x} \frac{\varphi(x)}{\psi(x)} d x .
\end{aligned}
$$

Thus,

$$
\int_{\mathbb{R}} \frac{\partial}{\partial t} \varphi(x) \ln \frac{\varphi(x)}{\psi(x)} d x=-\mathcal{I}_{a}(\varphi \| \psi)+\int_{\mathbb{R}} a(x) \psi(x) \frac{\partial^{2}}{\partial x^{2}} \frac{\varphi(x)}{\psi(x)}+b(x) \psi(x) \frac{\partial}{\partial x} \frac{\varphi(x)}{\psi(x)} d x
$$

Plugging this relation into equation 7.3 and take into account that $\psi(x)$ is also the solution of the Fokker Planck equation (7.1) we conclude that

$$
\frac{d}{d t} D_{K L}(\varphi \| \psi)=-\mathcal{I}_{a}(\varphi \| \psi)
$$

as desired.

## 8. Appendix

Here we give the proof of de Bruijn identity for a function running through a Gaussian channel.

Let $H(u)=\int_{\mathbb{R}} u(x, \cdot) \ln u(x, \cdot) d x$ be Shannon's entropy, then

$$
\begin{aligned}
\frac{d}{d t} H(u) & =\frac{d}{d t} \int_{\mathbb{R}} u(x, t) \ln u(x, t) d x \\
& \left.=\int_{\mathbb{R}}\left(\frac{d}{d t} u\right)(\ln u)+u \frac{d}{d t} \ln u(x, t)\right) d x \\
& =\int_{\mathbb{R}} \Delta u \ln u d x+\int_{\mathbb{R}} \frac{d}{d t} u(x, t) d x \\
& =-\int_{\mathbb{R}}\left((\nabla u)^{2} / u\right) d x+\frac{d}{d t} \int_{\mathbb{R}} u(x, t) d x
\end{aligned}
$$

Since $\int_{\mathbb{R}} u(x, t) d x=1$, we obtain $\frac{d}{d t} H(u)=\mathcal{I}(u)$.

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