Coupling time distribution asymptotics for some couplings of the Lévy stochastic area

Wilfrid S. Kendall

Abstract

We exhibit some explicit co-adapted couplings for \( n \)-dimensional Brownian motion and all its Lévy stochastic areas. In the two-dimensional case we show how to derive exact asymptotics for the coupling time under various mixed coupling strategies, using Dufresne’s formula for the distribution of exponential functionals of Brownian motion. This yields quantitative asymptotics for the distributions of random times required for certain simultaneous couplings of stochastic area and Brownian motion. The approach also applies to higher dimensions, but will then lead to upper and lower bounds rather than exact asymptotics.

Keywords Brownian motion, co-adapted coupling, coupling time distribution, Dufresne formula, exponential functional of Brownian motion, Kolmogorov diffusion, Lévy stochastic area, maximal coupling, Morse–Thue sequence, non-co-adapted coupling, reflection coupling, rotation coupling, stochastic differential, synchronous coupling

AMS subject classification (MSC2010) 60J65, 60H10

Introduction

It is a pleasure to present this paper as a homage to my DPhil supervisor John Kingman, in grateful acknowledgement of the formative period which I spent as his research student at Oxford, which launched me into

\[ \text{http://www.warwick.ac.uk/go/wsk} \]
W. S. Kendall

a deeply satisfying exploration of the world of mathematical research. It seems fitting in this paper to present an overview of a particular aspect of probabilistic coupling theory which has fascinated me for a considerable time; given that one can couple two copies of a random process, when can one in addition couple other associated functionals of the processes? How far can one go?

Motivations for this question include: the sheer intellectual curiosity of discovering just how far one can push the notion of probabilistic coupling; the consideration that coupling has established itself as an extremely powerful tool in probability theory and therefore that any increase in its scope is of potential significance; and the thought that the challenge of coupling in extreme circumstances may produce new paradigms in coupling to complement that of the classic reflection coupling.

It has been known since the 1970s that in principle one can couple two random processes maximally; at first encounter this fact continues to delight and surprise researchers. I summarize this point in Section 1 and also describe the important class of co-adapted couplings. These satisfy more restrictive requirements than maximal couplings, are typically less efficient, but are also typically much easier to construct. Since Lindvall (1982)’s seminal preprint we have known how to couple Euclidean Brownian motion using a simple reflection argument in a way which (most unusually) is both maximal and co-adapted, and this has led to many significant developments and generalizations, some of which are briefly sketched in Section 2. This leads to Section 3 which develops the main content of the paper; what can we now say about the question, how to couple not just Brownian motion, say, but also associated path integrals? Of course we then need to vary our strategy, using not just reflection coupling but also so-called synchronous coupling (in which the two processes move in parallel), and even rotation coupling, which correlates different coordinates of the two processes. In Kendall (2007) I showed how to couple (co-adaptively) $n$-dimensional Brownian motion and all its stochastic areas; this work is reviewed in Section 3.3 using a rather more explicit coupling strategy and then new computations are introduced (in Section 5.4) which establish explicit asymptotics for the coupling time for suitable coupling strategies in the two-dimensional case, and which can be used to derive naïve bounds in higher dimensions. Section 4 concludes the paper with some indications of future research directions.
1 Different kinds of couplings

Probabilistic couplings are used in many different ways: couplings (realizations of two random processes on the same probability space) can be constructed so as to arrange any of the following:

- for the two processes to agree after some random time (which is to say, for the coupling to be successful). This follows the pioneering work of Doeblin (1938), which uses this idea to provide a coupling proof of convergence to equilibrium for finite-state ergodic Markov chains;
- for the two processes to be interrelated by some monotonicity property—a common use of coupling in the study of interacting particle systems (Liggett, 2005);
- for one process to be linked to the other so as to provide some informative and fruitful representation, as in the case of the coalescent (Kingman, 1982);
- for one of the processes to be an illuminating approximation to the other; this appears in an unexpected way in Barbour et al. (1992)'s approach to Stein–Chen approximation.

These considerations often overlap. Aiming for successful coupling has historical precedence and is in some sense thematic for coupling theory, and we will focus on this task here.

1.1 Maximal couplings

The first natural question is, how fast can coupling occur? There is a remarkable and satisfying answer, namely that one can in principle construct a coupling which is maximal in the sense that it maximizes the probability of coupling before time $t$ for all possible $t$: see Griffeath (1974, 1978), Pitman (1976), Goldstein (1978). Briefly, maximal couplings convert the famous Aldous inequality (the probability of coupling is bounded above by a simple multiple of the total variation between distributions) into an equality. Constructions of maximal couplings are typically rather involved, and in general may be expected to involve demanding potential-theoretic questions quite as challenging as any problem which the coupling might be supposed to solve. Pitman’s approach is perhaps the most direct, involving explicit construction of suitable time-reversed Markov chains.
1.2 Co-adapted coupling

Maximal couplings being generally hard to construct, it is useful to consider couplings which are stricter in the sense of requiring the coupled processes both to be adapted to the same filtration. Terminology in the literature varies: *Markovian*, when the coupled processes are jointly Markov, with prescribed marginal kernels (Burdzy and Kendall, 2000); *co-immersed* ([Emery, 2005]) or *co-adapted* to emphasize the role of the filtration. The idea of a co-adapted coupling is simple enough, though its exact mathematical description is somewhat tedious: here we indicate the definition for Markov processes.

**Definition 1.1** Suppose one is given two continuous-time Markov processes $X^{(1)}$ and $X^{(2)}$, with corresponding semigroup kernels defined for bounded measurable functions $f$ by

$$P_t^{(i)} f(z) = E[f(X_{s+t}^{(i)}) \mid X_s^{(i)} = z, X_u^{(i)} \text{ for } u < s].$$

A co-adapted coupling of $X^{(1)}$ and $X^{(2)}$ is a pair of random processes $\tilde{X}^{(1)}$ and $\tilde{X}^{(2)}$ defined on the same filtered probability space $(\Omega, \mathcal{F}, \{\mathcal{F}_t : t \geq 0\}, P)$, both adapted to the common filtration $\{\mathcal{F}_t : t \geq 0\}$ (hence ‘co-adapted’) and satisfying

$$P_t^{(i)} f(z) = E[f(\tilde{X}_{s+t}^{(i)}) \mid \mathcal{F}_s, \tilde{X}_s^{(i)} = z]$$

for $i = 1, 2$, for each bounded measurable function $f$, each $z$, each $s$, $t > 0$.

Thus the individual stochastic dynamics of each $\tilde{X}^{(i)}$ agree with those of the corresponding $X^{(i)}$ even when the past behaviour of the opposite process is also taken into account. (This is typically not the case for maximal couplings.) In particular, if the $X^{(i)}$ are Brownian motions then their forward increments are independent of the past given by the filtration. Moreover if the processes are specified using stochastic differential equations driven by Brownian motion then general co-adapted couplings can be represented using stochastic calculus (possibly at the price of enriching the filtration), as observed in passing by [Emery, 2005], and as described more formally in [Kendall, 2009a, Lemma 6]. Briefly, any co-adapted coupling of vector-valued Brownian motions $A$ and $B$ can be represented by expressing $A$ as a stochastic integral with respect to $B$ and perhaps another independent Brownian motion $C$: we use this later at Equation (3.1).
1.3 Coupling at different times

There are many other useful couplings falling outside this framework: for example, Thorisson (1994) discusses the idea of a shift-coupling, which weakens the coupling requirement by permitting processes to couple at different times; Kendall (1994) uses co-adapted coupling of time-changed processes as part of an exploration of regularity for harmonic maps. However in this paper we will focus on co-adapted couplings.

2 Reflection coupling

The dominant example of coupling is reflection coupling for Euclidean Brownian motions $A$ and $B$, dating back to Lindvall (1982)'s preprint: construct $A$ from $B$ by reflecting $B$ in the line segment running from $B$ to $A$. That this is a maximal coupling follows from an easy computation involving the reflection principle. It can be expressed as a co-adapted coupling; the Brownian increment for $A$ is derived from that of $B$ by a reflection in the line segment running from $B$ to $A$. Many modifications of the reflection coupling have been derived to cover various situations; we provide a quick survey in the remainder of this section.

2.1 Maximality and non-maximality

The reflection coupling is unusual in being both co-adapted and maximal. Hsu and Sturm (2003) point out that reflection coupling fails to be maximal even for Euclidean Brownian motion if the Brownian motion is stopped on exit from a prescribed domain. (Kuwada and Sturm 2007 discuss the manifold case; see also Kuwada 2007, 2009.) Perhaps the simplest example of an instance where no co-adapted coupling can be maximal arises in the case of the Ornstein–Uhlenbeck process (Connor, 2007, PhD thesis, Theorem 3.15). Consider the problem of constructing successful co-adapted couplings between (i) an Ornstein–Uhlenbeck process begun at 0, and (ii) an Ornstein–Uhlenbeck process run in statistical equilibrium. A direct argument shows that no such co-adapted coupling can be maximal; however in this case reflection coupling is maximal amongst all co-adapted couplings. (The study of couplings which are maximal in the class of co-adapted couplings promises to be an interesting field: the case of random walk on the hypercube is treated by Connor and Jacka 2008.)
2.2 Coupling for Diffusions

A variant of reflection coupling for elliptic diffusions with smooth coefficients is discussed in Lindvall and Rogers (1986) and further in Chen and Li (1989). ‘Reflection’ here depends on interaction between the two diffusion matrices, and in general the two coupled diffusions do not play symmetrical roles. In the case of Brownian motion on a manifold one can use the mechanisms of stochastic development and stochastic parallel transport to define co-adapted couplings in a more symmetrical manner. The behaviour of general co-adapted Brownian couplings on Riemannian manifolds is related to curvature. Kendall (1986b) shows that successful co-adapted coupling can never be almost-surely successful in the case of a simply-connected manifold with negative curvatures bounded away from zero. On the other hand a geometric variant of reflection coupling known as mirror coupling will always be almost-surely successful if the manifold has non-negative Ricci curvatures (Kendall 1986a, Cranston 1992, Kendall 1988). Indeed Renesse (2004) shows how to generalize mirror coupling even to non-manifold contexts.

3 Coupling more than one feature of the process

The particular focus of this paper concerns ongoing work on the following question: is it possible co-adaptively to couple more than one feature of a random process at once? To be explicit, is it possible to couple not just the location but also some functional of the path?

On the face of it, this presents an intimidating challenge: control of difference of path functionals by coupling is necessarily indirect and it is possible that all attempts to control the discrepancy between path functionals will inevitably jeopardize coupling of the process itself.

Further thought shows that it is sensible to confine attention to cases where the process together with its functional form a hypoelliptic diffusion, since in such cases the Hörmander regularity theorem guarantees existence of a density, and this in turn shows that general coupling (not necessarily non-co-adapted) is possible in principle. (Hairer 2002 uses this approach to produce non-co-adapted couplings, using careful analytic estimates and a regeneration argument which corresponds to Lindvall 2002’s ‘γ coupling’.)

Furthermore it is then natural to restrict attention to diffusions with nilpotent group symmetries, where one may hope most easily to discover
co-adapted couplings which will be susceptible to extensive generalization, paralleling the generalizations of the Euclidean reflection coupling which have been described briefly in Section 2.

3.1 Kolmogorov diffusion

Consider the so-called Kolmogorov diffusion: scalar Brownian motion $B$ plus its time integral $\int B \, dt$. This determines a simple nilpotent diffusion, and in fact it can be coupled co-adaptively by varying sequentially between reflection coupling and synchronous coupling (allowing the two Brownian motions to move in parallel) as shown in Ben Arous et al. (1995). Jansons and Metcalfe (2007) describe some numerical investigations concerned with optimizing an exponential moment of the coupling time.

The idea underlying this coupling is rather simple. Suppose that we wish to couple $(B, \int B \, dt)$ with $(\tilde{B}, \int \tilde{B} \, dt)$. Set $U = \tilde{B} - B$ and $V = \int \tilde{B} \, dt - \int B \, dt$. Co-adapted couplings include stochastic integral representations such as $d\tilde{B} = J \, dB$, for co-adapted $J \in \{-1, 1\}$; $J = 1$ yields synchronous coupling and $J = -1$ yields reflected coupling. Suppose $U_0 \neq 0$ and $V = 0$ (which can always be achieved by starting with a little reflected or synchronous coupling unless $U = V = 0$ from the start, in which case nothing needs to be done). We can cause $(U, V)$ to wind repeatedly around $(0, 0)$ in ever smaller loops as follows: first use reflection coupling till $U$ hits $-U_0/2$, then synchronous coupling till $V$ hits $0$, then repeat the procedure but substituting $-U_0/2$ for $U_0$. A Borel–Cantelli argument combined with Brownian scaling shows that $(U, V)$ then hits $(0, 0)$ in finite time.

Kendall and Price (2004) present a cleaned-up version of this argument (together with an extension to deal in addition with $\int \int B \, ds \, dt$).

Curiously, this apparently artificial example can actually be applied to the study of the tail $\sigma$-algebra of a certain relativistic diffusion discussed by Bailleul (2008).

Remarkably it is possible to do very much better by using a completely different and implicit method: one can couple not just the time integral, but also any finite number of additional iterated time integrals (Kendall and Price, 2004), by concatenating reflection and synchronous couplings using the celebrated Morse–Thue binary sequence 0110100110010110 . . .

Scaled iterations of state-dependent perturbations of the resulting concatenation of couplings can be used to deliver coupling in a finite time;
the perturbed Morse–Thue sequences encode indirect controls of higher-order iterated integrals.

### 3.2 Lévy stochastic areas

Moving from scalar to planar Brownian motion, the natural question is now whether one can co-adaptively couple the nilpotent diffusion formed by Brownian motion \((B_1, B_2)\) and the Lévy stochastic area \(\int (B_1 \, dB_2 - B_2 \, dB_1)\). This corresponds to coupling a hypoelliptic Brownian motion on the Heisenberg group, and Ben Arous et al. determine an explicit successful coupling based on extensive explorations using computer algebra.

Again one can do better [Kendall, 2007]. Not only can one construct a simplified coupling for the 2-dimensional case based only on reflection and synchronous couplings (switching from reflection to synchronous coupling whenever a geometric difference of the stochastic areas exceeds a fixed multiple of the squared distance between the two coupled Brownian motions), but also one can successfully couple \(n\)-dimensional Brownian motion plus a \(\binom{n}{2}\) basis of the various stochastic areas. In the remainder of this paper we will indicate the method used, which moves beyond the use of reflection and synchronous couplings to involve rotation couplings as well (in which coordinates of one of the Brownian motions can be correlated to quite different coordinates of the other).

### 3.3 Explicit strategies for coupling Lévy stochastic area

Here we describe a variant coupling strategy for the \(n\)-dimensional case which is more explicit than the strategy proposed in Kendall (2007).

As described in Kendall (2007, Lemma 6), suppose that \(\pmb{A}\) and \(\pmb{B}\) are co-adaptively coupled \(n\)-dimensional Brownian motions. Arguing as in Kendall (2009a), and enriching the filtration if necessary, we may represent any such coupling in terms of a further \(n\)-dimensional Brownian motion \(\pmb{C}\), independent of \(\pmb{B}\):

\[
\frac{d\pmb{A}}{d\pmb{B}} = \frac{\frac{d\pmb{B}}{d\pmb{C}} + \frac{\frac{d\pmb{C}}{d\pmb{C}}}{2}}{d\pmb{C}},
\]

(3.1)

where \(\frac{d\pmb{B}}{d\pmb{C}}\) are predictable \((n \times n)\)-matrix-valued processes satisfying the constraint

\[
\frac{d\pmb{B}}{d\pmb{C}} + \frac{d\pmb{C}}{d\pmb{C}} = \frac{1}{2}
\]

(3.2)

and where \(1\) represents the \((n \times n)\) identity matrix.
Note that the condition \((3.2)\) is equivalent to the following set of symmetric matrix inequalities for the co-adapted process \(J\) (interpreted in a spectral sense):
\[
J^\top J \leq I.
\]
(3.3)

Thus a legitimate coupling control \(J\) must take values in a compact convex set of \(n \times n\) matrices defined by (3.3).

The matrix process \(J\) measures the correlation \((dB_i dB_j dni) / dt\) between the Brownian differentials \(dB_i\) and \(dB_j\): for convenience let \(S = \frac{1}{2}(J + J^\top)\) and \(A = \frac{1}{2}(J - J^\top)\) refer to the symmetric and skew-symmetric parts of \(J\). The coupling problem solved in Kendall (2007) is to choose an adapted \(J = S + A\) which brings \(A\) and \(B\) into agreement at a coupling time \(T_{\text{coupling}}\) which is simultaneously a coupling time for all the \(\binom{n}{2}\) corresponding pairs of stochastic area integrals \(\int (A_i dA_j - A_j dA_i)\) and \(\int (B_i dB_j - B_j dB_i)\).

To measure progress towards this simultaneous coupling, set \(X = A - B\) and define \(\mathfrak{A}\) to be a skew-symmetric matrix of geometric differences between stochastic areas with \((i,j)^{th}\) entry
\[
\mathfrak{A}_{ij} = \int (A_i dA_j - A_j dA_i) - \int (B_i dB_j - B_j dB_i) + A_i B_j - A_j B_i.
\]
(3.4)

The nonlinear correction term \(A_i B_j - A_j B_i\) is important because it converts \(\mathfrak{A}_{ij}\) into a geometrically natural quantity, invariant under shifts of coordinate system, and also because it supplies a very useful contribution to the drift in the subsequent Itô analysis. Of course \(\mathfrak{A}\) and \(X\) both vanish at a given time \(t\) if and only if at that time both \(A = B\) (so in particular the correction term vanishes) and also all the corresponding stochastic areas agree.

Some detailed Itô calculus (originally carried out in an implementation of the Itô calculus procedures ‘Rouvs’ in Axiom, Kendall, 2001, but now checked comprehensively by hand) can now be used to derive the following system of stochastic differential equations for the squared distance \(V^2 = \|X\|^2\) and the ‘squared areal difference’ \(U^2 = \text{trace}(\mathfrak{A}^\top \mathfrak{A}) = \sum \sum_{ij} \mathfrak{A}_{ij}^2\):

\[
(d(V^2))^2 = 8 \nu^\top (\mathbb{1} - S) \nu V^2 \, dt,
\]

Drift \(d(V^2) = 2 \text{trace}(\mathbb{1} - S) \nu \, dt\),

\[
d(V^2) \times d(U^2) = -16 \nu^\top Z^\top A \nu UV^2 \, dt,
\]

\[
(d(U^2))^2 = 32 \nu^\top Z^\top (\mathbb{1} + S) Z \nu U^2 V^2 \, dt,
\]

\[
(d(U^2))^2 = 32 \nu^\top Z^\top (\mathbb{1} + S) Z \nu U^2 V^2 \, dt.
\]
\[
\text{Drift } d(U^2) = 4 \text{trace} \left( Z^\top A \right) U \, dt + \\
+ 4 \left( \text{trace} \left( \mathbb{1} + S \right) - \nu^\top (\mathbb{1} + S) \nu \right) V^2 \, dt.
\]

(3.5)

Here the vector \( \nu \) and the matrix \( Z \) encode relevant underlying geometry: respectively \( V \nu = X \) and \( UZ = A \). Note that \( \nu \) is a unit vector and \( Z \) has unit Hilbert–Schmidt norm: \( \text{trace} Z^\top Z = 1 \).

The strategy is to consider \( V^2 \) and \( U^2 \) on a log-scale: further stochastic calculus together with suitable choice of coupling control \( J \) then permits comparison to two Brownian motions with constant negative drift in a new time-scale, and a stochastic-calculus argument shows that \( K = \frac{1}{2} \log(V^2) \) and \( H = \frac{1}{2} \log(U^2) \) reach \(-\infty\) at finite time in the original time-scale measured by \( t \). In fact further stochastic calculus, based on the martingale differential identity

\[
d \log Z = \frac{dZ}{Z} - \frac{1}{2} \left( \frac{dZ}{Z} \right)^2,
\]

shows that

\[
(\text{d}K)^2 = \frac{1}{2} \nu^\top (\mathbb{1} - S) \nu \, d\tilde{\tau},
\]

\[
\text{Drift}(\text{d}K) = \frac{1}{4} \left( \text{trace} (\mathbb{1} - S) - 2 \nu^\top (\mathbb{1} - S) \nu \right) \, d\tilde{\tau},
\]

\[
dK \times dH = -\nu^\top Z^\top A \nu \frac{d\tilde{\tau}}{W},
\]

\[
(\text{d}H)^2 = 2 \nu^\top Z^\top (\mathbb{1} + S) \nu \, d\tilde{\tau},
\]

\[
\text{Drift}(\text{d}H) = \frac{1}{2} \text{trace} \left( Z^\top A \right) \frac{d\tilde{\tau}}{W} + \\
+ \frac{1}{2} \left( \text{trace} (\mathbb{1} + S) - \nu^\top (\mathbb{1} + S) \nu - 4 \nu^\top Z^\top (\mathbb{1} + S) Z \nu \right) \frac{d\tilde{\tau}}{W^2}.
\]

(3.6)

Here \( d\tilde{\tau} = 4 \frac{dt}{V^2} \) determines the new time-scale \( \tilde{\tau} \), and \( W = U/V^2 = \exp(H - 2K) \). It is clear from the system (3.6) that the contribution of \( \frac{1}{2} \text{trace} \left( Z^\top A \right) \frac{d\tilde{\tau}}{W} \) (deriving ultimately from the areal difference correction term in (3.4)) is potentially a flexible and effective component of the control if \( 1/W^2 \) is small. On the other hand if this is to be useful then \( \mathbb{1} - S \) must be correspondingly reduced so as to ensure that \( H \) and \( K \) are subject to dynamics on comparable time-scales.

Kendall (2007) considers the effect of a control \( J \) which is an affine mixture of reflection \( \mathbb{1} - 2\nu \nu^\top \) and rotation \( \exp(-\theta Z) \). Second-order Taylor series expansion of the matrix exponential is used to overcome analytical complexities at the price of some mild asymptotic analysis.
Here we indicate an alternative route, replacing the second-order truncated expansion $J' = I - \theta ZZ^T - \frac{\theta^2}{2} ZZ^T Z$ (which itself fails to satisfy (3.3) and is not therefore a valid coupling control) by $J'' = I - \theta Z - \theta^2 ZZ^T Z$, which does satisfy (3.3) when $\theta^2 \leq 2$ (using the fact that non-zero eigenvalues of $ZZ^T$ have multiplicity 2, so that Trace($ZZ^T$) = 1 implies that $0 \leq ZZ^T \leq \frac{1}{2} I$). Thus we can consider

$$J = I - 2p \nu \nu^T - (1-p)\theta Z - (1-p)\theta^2 ZZ^T Z,$$

which is a valid coupling control satisfying (3.3) when $0 \leq p \leq 1$ and $\theta^2 \leq 2$.

This leads to the following stochastic differential system, where terms which will eventually be negligible have been separated out:

$$(dK)^2 = \left( p + \frac{1}{2} \|Z\nu\|^2 \theta^2 \right) d\bar{\tau} - \frac{\theta}{2} \|Z\nu\|^2 \theta^2 \ d\bar{\tau},$$

Drift($dK$) = $-\frac{1}{2} \left( p - \frac{1}{2} (1 - 2\|Z\nu\|^2) \theta^2 \right) d\bar{\tau}$

$+ \frac{\theta}{4} (1 - 2\|Z\nu\|^2) \theta^2 \ d\bar{\tau},$

$dK \times dH = \|Z\nu\|^2 \theta \frac{d\tau}{\tau} - p\|Z\nu\|^2 \theta \frac{d\tau}{\tau},$

$$(dH)^2 = 4 \|Z\nu\|^2 \frac{d\tau}{\tau} - 2(1-p)\|Z^T Z\nu\|^2 \theta^2 \frac{d\tau}{\tau},$$

Drift($dH$) = $-\frac{1}{2} \theta \frac{d\tau}{\tau} + (n - 4\|Z\nu\|^2) \frac{d\tau}{\tau}$

$+ \frac{1}{2} p \theta \frac{d\tau}{\tau} - \frac{1-p}{2} \left( 1 - \|Z\nu\|^2 - 4 \|Z^T Z\nu\|^2 \right) \theta^2 \frac{d\tau}{\tau}.$

(3.8)

In order to ensure comparable dynamics for $K$ and $H$, set $p = \alpha^2 / W^2$ and $\theta = \beta / W$ (valid when $W \geq \max \{ \frac{1}{\sqrt{2}} \beta, \alpha \}$): writing $d\tau = d\bar{\tau} / W^2 = 4(V/U)^2 \ d t$ leads to

$$(dK)^2 = \left( \alpha^2 + \frac{1}{2} \|Z\nu\|^2 \beta^2 - \frac{1}{2} \|Z\nu\|^2 \alpha^2 \beta^2 \right) \ d\tau,$$

Drift($dK$) = $-\mu_1 \ d\tau,$

d$K \times dH = \|Z\nu\|^2 \left( 1 - \frac{\alpha^2}{\tau^2} \right) \beta \ d\tau,$

$$(dH)^2 = 4 \left( \|Z\nu\|^2 - \frac{1}{2}(1 - \frac{\alpha^2}{\tau^2}) \|Z^T Z\nu\|^2 \beta^2 \right) \ d\tau,$$

Drift($dH$) = $-\mu_2 \ d\tau,$

(3.9)
where $\mu_1$ and $\mu_2$ are given by

$$\mu_1 = \frac{1}{2} \alpha^2 - \frac{1}{4} (1 - 2\|Z\nu\|^2) \beta^2 + \frac{1}{4\pi^2} (1 - 2\|Z\nu\|^2) \alpha^2 \beta^2,$$

$$\mu_2 = \frac{1}{2} \beta - n + 1 + 4\|Z\nu\|^2 - \frac{1}{2} W^2 \left( \alpha^2 \beta - (1 - \frac{\alpha^2}{\pi^2}) \left( 1 - \|Z\nu\|^2 - 4 \|Z^\top Z\nu\|^2 \right) \beta^2 \right).$$

(3.10)

In order to fulfil the underlying strategy, $\mu_1$ and $\mu_2$ should be chosen to accomplish the following:

1. $W = \exp(H - 2K)$ must remain large; this follows by a strong-law-of-large-numbers argument if $2\mu_1 - \mu_2 = \text{Drift } d(H - 2K)/d\tau$ is positive and bounded away from zero and yet $(dH)^2/d\tau$ and $(dK)^2/d\tau$ are bounded.

2. Both of $-\mu_1 = \text{Drift } d(K)/d\tau$ and $-\mu_2 = \text{Drift } d(H)/d\tau$ must remain negative and bounded away from zero, so that $K$ and $H$ both tend to $-\infty$ as $\tau \to \infty$.

3. If coupling is to happen in finite time on the $t$-time-scale then

$$T_{\text{coupling}} = \frac{1}{4} \int_0^\infty \left( \frac{U}{V} \right)^2 d\tau = \frac{1}{4} \int_0^\infty \exp(2H - 2K) d\tau < \infty.$$ This follows almost surely if $\mu_1, \mu_2$ are chosen so that $2\mu_1 - 2\mu_2 = \text{Drift } d(2H - 2K)/d\tau$ is negative and bounded away from zero.

Now the inverse function theorem can be applied to show that for any constant $L > 0$ there is a constant $L' > 0$ such that if $W^2 \geq L'$ then (3.10) can be solved for any prescribed $0 \leq \mu_1, \mu_2 \leq L$ using $\alpha^2, \beta^2 \leq L'$ (incidentally thus bounding $(dH)^2/d\tau$ and $(dK)^2/d\tau$). Moreover by choosing $L'$ large enough it then follows that $p = \alpha^2/W^2 \leq 1$ and $\beta^2/W^2 \leq 2$, so that the desired $\mu_1, \mu_2$ can be attained using a valid coupling control.

A comparison with Brownian motion of constant drift now shows that if initially $W \geq 2L'$ then there is a positive chance that $W \geq L'$ for all $\tau$, and thus that coupling happens at $\tau$-time infinity, and actual time $T_{\text{coupling}} < \infty$. Should $W$ drop below $L'$, then one can switch to the pure reflection control ($p = 1$) and run this control until $W = U/V^2$ rises again to level $2L'$. This is almost sure to happen eventually, since otherwise $V \to 0$ and thus $U = WV^2 \to 0$, which can be shown to have probability 0 under this control.
3.4 Estimates for coupling time distribution

In the planar case \( n = 2 \) the stochastic differential system (3.6) under mixed rotation-reflection controls can be simplified substantially. In this section we go beyond the work of Ben Arous et al. (1995) and Kendall (2007) by using this simplification to identify limiting distributions for suitable scalings of the coupling time \( T_{\text{coupling}} \). The simplification arises because the non-zero eigenvalues of the skew-symmetric matrix \( Z \) have even multiplicity; consequently in the two-dimensional case we may deduce from \( \text{trace}(Z^T Z) = 1 \) and \( \|\nu\| = 1 \) that

\[
\|Z\nu\|^2 = 1
\]

and

\[
\|Z^T Z\nu\|^2 = \frac{1}{4}.
\]

Moreover an Euler formula follows from

\[
Z Z = -\frac{1}{2}I;
\]

\[
\exp(-\sqrt{2} \theta \, Z) = \cos \theta I - \sqrt{2} \sin \theta \, Z.
\]

Accordingly in the planar case the mixed coupling control

\[
J = p (I - 2 \nu \nu^T) + q \exp(-\sqrt{2} \theta \, Z)
\]

(3.11)

renders the system (3.6) as

\[
\begin{align*}
(\, dK \,)^2 & = \min\{W^2, \alpha^2\} \, d\tau, \\
\text{Drift}(\, dK \,) & = -\frac{1}{2}p \, d\tau, \\
dK \times dH & = 0, \\
(\, dH \,)^2 & = 2 \, d\tau, \\
\text{Drift}(\, dH \,) & = -d\tau.
\end{align*}
\]

(3.12)

If we set \( \theta = 0 \) (so that the coupling control is a mixture of reflection coupling and synchronous couplings) then the result can be made to yield a successful coupling: choosing \( p = \min\{1, \alpha^2/W^2\} \) the stochastic differential system becomes

\[
\begin{align*}
(\, dK \,)^2 & = \min\{W^2, \alpha^2\} \, d\tau, \\
\text{Drift}(\, dK \,) & = -\frac{1}{2} \min\{W^2, \alpha^2\} \, d\tau, \\
dK \times dH & = 0, \\
(\, dH \,)^2 & = 2 \, d\tau, \\
\text{Drift}(\, dH \,) & = -d\tau.
\end{align*}
\]

(3.13)

where \( d\tau = 4 \, dt/V^2 \) as before. Accordingly, for fixed \( \alpha^2 \), once \( W^2 \geq \alpha^2 \) then \( H \) and \( K \) behave as uncorrelated Brownian motions with constant
negative drifts in the $\tau$ time-scale; moreover if $\alpha^2 > 1$ then
\[
\text{Drift } \frac{d(\log W)}{d\tau} = \alpha^2 - 1
\]
is strictly positive and so $W \to \infty$ almost surely. In order to argue as before we must finally show almost-sure finiteness of
\[
T_{\text{coupling}} = \frac{1}{4} \int_0^\infty \exp (2H - 2K) \, d\tau.
\]
Now if we scale using the ratio $(U_0/V_0)^2$ at time zero then we can deduce the following convergence-in-distribution result as $W_0 = U_0/V_0^2 \to \infty$:
\[
\left( \frac{V_0}{U_0} \right)^2 T_{\text{coupling}} = \frac{1}{4} \int_0^\infty \exp (2H - 2K - (2H_0 - 2K_0)) \, d\tau \to \frac{1}{4} \int_0^\infty \exp \left( 2\sqrt{2 + \alpha^2} \bar{B} - (2 - \alpha^2)\tau \right) \, d\tau
\]
(3.14)
where $\bar{B}$ is a standard real Brownian motion begun at 0. This integral is finite when $\alpha^2 < 2$, so finally we deduce that coupling occurs in finite time for this simple mixture of reflection and synchronous coupling if we choose $1 < \alpha^2 < 2$.

However we can now say much more, since the stochastic integral in (3.14) is one of the celebrated and much-studied exponential functionals of Brownian motion (Yor, 1992, or Yor, 2001, p. 15). In particular Dufresne (1990) has shown that such a functional
\[
\int_0^\infty \exp \left( a\bar{B}_s - bs \right) \, ds
\]
has the distribution of $2/(a^2 \Gamma_{2b/a^2})$, where $\Gamma_\kappa$ is a Gamma-distributed random variable of index $\kappa$. In summary,

**Theorem 3.1** Let $T_{\text{coupling}}$ be the coupling time for two-dimensional Brownian motion plus Lévy stochastic area under a mixture of reflection coupling and synchronous coupling. Using the notation above, let $\min\{1, \alpha^2/W^2\}$ be the proportion of reflection coupling. Scale $T_{\text{coupling}}$ by the square of the ratio between initial areal difference $U_0$ and initial spatial distance $V_0$; if $1 < \alpha^2 < 2$ then as $W_0 = U_0/V_0^2 \to \infty$ so a re-scaling of the coupling time has limiting Inverse Gamma distribution
\[
\left( \frac{V_0}{U_0} \right)^2 T_{\text{coupling}} \to \frac{2}{2 + \alpha^2 \Gamma_{(2-\alpha^2)/(2(2+\alpha^2))}}.
\]
Thus the tail behaviour of $T_{\text{coupling}}$ is governed by the index of the Gamma random variable, which here cannot exceed $\frac{1}{\alpha}$ (the limiting case when $\alpha^2 \to 1$). This index is unattainable by this means since $H = 2K$ behaves like a Brownian motion when $\alpha^2 = 1$, so we cannot have $W \to \infty$. Kendall (2007, Section 3) exhibits a similar coupling for the planar case in which there is a state-dependent switching between reflection and synchronous coupling, depending on whether the ratio $W$ exceeds a specified threshold; it would be interesting to calculate the tail-behaviour of the inverse of the coupling time in this case.

Note that scaling by $(U_0/V_0)^2$ rather than $W_0 = U_0/V_0^2$ quantifies something which can be observed from detailed inspection of the stochastic differential system (3.5); the rate of evolution of the areal distance $U$ is reduced if the spatial distance $V$ is small. The requirement $W_0 \to \infty$ is present mainly to remove the effect of higher-order terms $(d\tau/W^2)$ in systems such as (3.9), and in particular to ensure in (3.13) that $\min\{W^2, \alpha^2\} = \alpha^2$ for all time with high probability.

In fact one can do markedly better than the reflection-synchronous mixture coupling of Theorem 3.1 by replacing synchronous coupling by a rotation coupling for which $\sin \theta = \sqrt{2} \beta/W$: similar calculations then show the index of the inverse of the limiting scaled coupling time can be increased up to the limit of $\frac{1}{4}$, which remarkably is the index of the inverse of the coupling time for reflection coupling of Brownian motion alone! However this limit is not attainable by these mixture couplings, as it corresponds to a limiting case of $\beta = 2$, $\alpha^2 = 3$. At this choice of parameter values $H = 2K$ again behaves like a Brownian motion so we do not have $W \to \infty$.

In higher dimensions similar calculations can be carried out, but the geometry is more complicated; in the planar case the form of $Z$ is essentially constant, whereas it will evolve stochastically in higher dimensions and relate non-trivially to $\nu$. This leads to correspondingly weaker results: the inverse limiting scaled coupling time can be bounded above and below using Gamma distributions of different indices.

We should not expect these mixture couplings to be maximal, even within the class of co-adapted couplings. Indeed Kendall (2007) gives a heuristic argument to show that maximality amongst co-adapted couplings should be expected only when one Brownian differential is a (state-dependent) rotation or rotated reflection of the other. The interest of these mixture couplings lies in the ease with which one may derive limiting distributions for them, hence gaining a good perspective on how rapidly one may couple the stochastic area.
4 Conclusion

After reviewing aspects of coupling theory, we have indicated an approach to co-adapted coupling of Brownian motion and its stochastic areas, and shown how in the planar case one can use Dufresne’s formula to derive asymptotics of coupling time distributions for suitable mixed couplings. Aspects of these asymptotic distributions indicate the price that is to be paid for coupling stochastic areas as well as the Brownian motions themselves; however it is clear that these mixed couplings should not be expected to be maximal amongst all co-adapted couplings. Accordingly a very interesting direction for future research is to develop these methods to derive estimates for coupling time distributions for more efficient couplings using state-dependent coupling strategies as exemplified in Kendall (2007, Section 3). Progress in this direction would deliver probabilistic gradient estimates in the manner of Cranston (1991, 1992) (contrast the analytic work of Bakry et al. 2008).

A further challenge is to develop these techniques for higher-dimensional cases. Here the two-dimensional approach extends naively to deliver upper and lower bounding distributions; a more satisfactory answer with tighter bounds will require careful analysis of the evolution under the coupling of the geometry as expressed by the pair $(\nu, Z)$.

A major piece of unfinished business in this area is to determine the extent to which these co-adapted coupling results extend to higher-order iterated path integrals (simple Itô calculus demonstrates that it suffices to couple Lévy stochastic areas in order to couple all possible non-iterated path integrals of the form $\int B_i \, dB_j$). Some tentative insight is offered by the rôle played by the Morse–Thue sequence for iterated time-integrals (Kendall and Price 2004). Moreover it is possible to generalize the invariance considerations underlying (5.3) for the areal difference, so as to produce similarly invariant differences of higher-order iterated path integrals. But at present the closing question of Kendall (2007) still remains open, whether one can co-adaptively couple Brownian motions together with all possible iterated path and time-integrals up to a fixed order of iteration.

References

Bailleul, I. 2008. Poisson boundary of a relativistic diffusion. Probab. Theory Related Fields, 141(1-2), 283–329.

Bakry, D., Baudoin, F., Bonnefont, M., and Chafaï, D. 2008. On gradient
bounds for the heat kernel on the Heisenberg group. *J. Funct. Anal.*, 255(8), 1905–1938.

Barbour, A. D., Holst, L., and Janson, S. 1992. *Poisson Approximation*. Oxford Studies in Probability, vol. 2. New York: Oxford Univ. Press. Oxford Science Publications.

Ben Arous, G., Cranston, M., and Kendall, W. S. 1995. Coupling constructions for hypoelliptic diffusions: two examples. Pages 193–212 of: Cranston, M., and Pinsky, M. (eds), *Stochastic Analysis: Summer Research Institute July 11-30, 1993*. Proc. Sympos. Pure Math., vol. 57. Providence, RI: Amer. Math. Soc.

Burdzy, K., and Kendall, W. S. 2000. Efficient Markovian couplings: examples and counterexamples. *Ann. Appl. Probab.*, 10(2), 362–409.

Chen, M. F., and Li, S. F. 1989. Coupling methods for multidimensional diffusion processes. *Ann. Probab.*, 17(1), 151–177.

Connor, S. B. 2007. *Coupling: Cutoffs, CFTP and Tameness*. Ph.D. thesis, Department of Statistics, University of Warwick.

Connor, S. B., and Jacka, S. D. 2008. Optimal co-adapted coupling for the symmetric random walk on the hypercube. *J. Appl. Probab.*, 45(1), 703–713.

Cranston, M. 1991. Gradient estimates on manifolds using coupling. *J. Funct. Anal.*, 99(1), 110–124.

Cranston, M. 1992. A probabilistic approach to gradient estimates. *Canad. Math. Bull.*, 35(1), 46–55.

Doeblin, W. 1938. Exposé de la théorie des chaînes simples constants de Markoff à un nombre fini d’états. *Revue Math. de l’Union Interbalkanique*, 2, 77–105.

Dufresne, D. 1990. The distribution of a perpetuity, with applications to risk theory and pension funding. *Scand. Actuar. J.*, (1-2), 39–79.

Émery, M. 2005. On certain almost Brownian filtrations. *Ann. Inst. H. Poincaré Probab. Statist.*, 41(3), 285–305.

Goldstein, S. 1978. Maximal coupling. *Z Wahrscheinlichkeitstheorie verw. Gebiete*, 46(2), 193–204.

Griffeath, D. 1974. A maximal coupling for Markov chains. *Z Wahrscheinlichkeitstheorie verw. Gebiete*, 31, 95–106.

Griffeath, D. 1978. Coupling methods for Markov processes. Pages 1–43 of: *Studies in Probability and Ergodic Theory*. Adv. in Math. Suppl. Stud., vol. 2. New York: Academic Press.

Hairer, M. 2002. Exponential mixing properties of stochastic PDEs through asymptotic coupling. *Probab. Theory Related Fields*, 124(3), 345–380.

Hsu, E. P., and Sturm, K.-T. 2003. *Maximal Coupling of Euclidean Brownian Motions*. SFB-Preprint 85. Universität Bonn.

Jansons, K. M., and Metcalfe, P. D. 2007. Optimally coupling the Kolmogorov diffusion, and related optimal control problems. *J. Comput. Math.*, 10, 1–20.

Kendall, W. S. 1986a. Nonnegative Ricci curvature and the Brownian coupling property. *Stochastics and Stochastic Reports*, 19, 111–129.
Kendall, W. S. 1986b. Stochastic differential geometry, a coupling property, and harmonic maps. *J. Lond. Math. Soc.* (2), **33**, 554–566.

Kendall, W. S. 1988. Martingales on manifolds and harmonic maps. Pages 121–157 of: Durrett, R., and Pinsky, M. (eds), *The Geometry of Random Motion*. Contemp. Math., vol. 73. Providence, RI: Amer. Math. Soc.

Kendall, W. S. 1994. Probability, convexity, and harmonic maps II: Smoothness via probabilistic gradient inequalities. *J. Funct. Anal.*, **126**, 228–257.

Kendall, W. S. 2001. Symbolic Itô calculus: an ongoing story. *Stat. Comput.*, **11**, 25–35.

Kendall, W. S. 2007. Coupling all the Lévy stochastic areas of multidimensional Brownian motion. *Ann. Probab.*, **35**(3), 935–953.

Kendall, W. S. 2009a. Brownian couplings, convexity, and shy-ness. *Electron. Commun. Probab.*, **14**, 63–72.

Kuwada, K. 2009. Characterization of maximal Markovian couplings for diffusion processes. *Electron. J. Probab.*, **14**, 63–72.

Liggett, T. M. 2005. *Interacting Particle Systems*. Classics Math. Berlin: Springer-Verlag. Reprint of the 1985 original.

Lindvall, T. 1982. *On Coupling of Brownian Motions*. Tech. rept. 1982:23. Department of Mathematics, Chalmers University of Technology and University of Göteborg.

Lindvall, T. 2002. *Lectures on the Coupling Method*. Mineola, NY: Dover Publications. Corrected reprint of the 1992 original.

Lindvall, T., and Rogers, L. C. G. 1986. Coupling of multidimensional diffusions by reflection. *Ann. Probab.*, **14**(3), 860–872.

Pitman, J. 1976. On coupling of Markov chains. *Z Wahrscheinlichkeitstheorie verw. Gebiete.*, **35**(4), 315–322.

Renéssé, M. von. 2004. Intrinsic coupling on Riemannian manifolds and polyhedra. *Electron. J. Probab.*, **9**, 411–435.

Thorisson, H. 1994. Shift-coupling in continuous time. *Probab. Theory Related Fields*, **99**(4), 477–483.

Yor, M. 1992. Sur certaines fonctionnelles exponentielles du mouvement brownien réel. *J. Appl. Probab.*, **29**(1), 202–208.

Yor, M. 2001. *Exponential Functionals of Brownian Motion and Related Processes*. Springer Finance. Berlin: Springer-Verlag.