## 4 More on Brownian rotations

#### 4a Moment method on the circle

We return to the process  $Y_t = \exp(i\sigma B_t + ivt)$ ,  $\sigma, v \in \mathbb{R}$  (as in 3a23). The distribution of  $Y_t$  is easy to describe explicitly by a density: for  $0 < u < v < 2\pi$ ,

(4a1) 
$$\mathbb{P}\left(Y_t \in e^{i[u,v]}\right) = \int_u^v p_t(x) \, dx,$$

$$p_t(x) = \frac{1}{\sqrt{2\pi}|\sigma|\sqrt{t}} \sum_{k=-\infty}^{+\infty} \exp\left(-\frac{(x-vt+2\pi k)^2}{2\sigma^2 t}\right)$$

(unless  $\sigma = 0$ , of course); here  $e^{i[u,v]}$  denotes the arc  $\{e^{ix} : u \leq x \leq v\}$ . This approach, however, does not work for more general situations of 3c, when  $Y_t$  may depend on the past of  $(X_s)$ . This is why we turn to the moment method.

For any random variable  $U:\Omega\to\mathbb{T}=\{z\in\mathbb{C}:|z|=1\}$ , the distribution  $\mu$  of U is uniquely determined by its moments

$$\mathbb{E} U^k = \int_{\mathbb{T}} u^k \, \mu(du) \in \mathbb{C}, \qquad k = 1, 2, \dots$$

since functions 1,  $\cos kx$ ,  $\sin kx$  span the space of continuous  $2\pi$ -periodic functions.

For any  $k = 1, 2, \ldots$  the process  $(Y_t^k)_t$  is another Brownian motion in  $\mathbb{T}$ , and

$$\mathbb{E}Y_t^k = \mathbb{E}\exp(ik\sigma B_t + ikvt) = \exp\left(-\frac{1}{2}k^2\sigma^2t\right)e^{ikvt} = \exp\left(\left(-\frac{1}{2}k^2\sigma^2 + ikv\right)t\right).$$

By the way, for  $t \to \infty$  we get  $\mathbb{E}Y_t^k \to 0$  (unless  $\sigma = 0$ ), which means that the distribution converges (weakly) to the uniform distribution on  $\mathbb{T}$ . (Do you see it via (4a1)?)

#### 4b Tensor moments of random matrices

**4b1 Exercise.** The distribution of a random matrix  $U: \Omega \to SO(n)$ , in general, is not determined uniquely by the matrices  $\mathbb{E}U^k$ ,  $k=1,2,\ldots$ 

Prove it (by finding a counterexample).

Hint: restrict yourself to diagonal matrices.

**4b2 Definition.** For any matrices  $A \in \mathrm{M}_m(\mathbb{R})$ ,  $B \in \mathrm{M}_n(\mathbb{R})$ , their tensor product  $A \otimes B \in \mathrm{M}_{mn}(\mathbb{R})$  is

$$(A \otimes B)_{\alpha,\beta}^{\gamma,\delta} = A_{\alpha}^{\gamma} B_{\beta}^{\delta} \quad \text{for } \alpha, \gamma \in \{1, \dots, m\} \text{ and } \beta, \delta \in \{1, \dots, n\}.$$

You see, rows and columns of  $A \otimes B$  are numbered by  $\{1, \ldots, m\} \times \{1, \ldots, n\}$  rather than  $\{1, \ldots, mn\}$ . The freedom of enumerating the pairs does not influence algebraic relations between matrices. An example:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \otimes \begin{pmatrix} e & f \\ g & h \end{pmatrix} = \begin{pmatrix} ae & af & be & bf \\ ag & ah & bg & bh \\ ce & cf & de & df \\ cg & ch & dg & dh \end{pmatrix}$$

(which enumeration is used here?).

Tensor product of several matrices is defined similarly. In particular,

$$A^{\otimes k} = \underbrace{A \otimes \cdots \otimes A}_{k}.$$

**4b3 Exercise.** For any random matrix  $U: \Omega \to SO(n)$ , the distribution  $\mu$  of U is uniquely determined by its tensor moments

$$\mathbb{E} U^{\otimes k} = \int_{SO(n)} u^{\otimes k} \, \mu(du) \in \mathcal{M}_{n^k}(\mathbb{R}) \,.$$

Prove it.

Hint: polynomials are dense among continuous functions on SO(n).

**4b4 Exercise.** (a)  $(A \otimes B)(x \otimes y) = (Ax) \otimes (By)$  for all  $x \in \mathbb{R}^m$ ,  $y \in \mathbb{R}^n$ ,  $A \in M_m(\mathbb{R})$ ,  $B \in M_n(\mathbb{R})$ . Here  $x \otimes y \in \mathbb{R}^{mn}$  is defined by

$$(x \otimes y)_{\alpha,\beta} = x_{\alpha}y_{\beta}$$
 for  $\alpha \in \{1, \dots, m\}, \beta \in \{1, \dots, n\}$ 

(up to enumeration...). Similarly,  $(x \otimes y)(A \otimes B) = (xA) \otimes (yB)$ .

(b)  $(A \otimes B)(C \otimes D) = (AC) \otimes (BD)$  for all  $A, C \in M_m(\mathbb{R}), B, D \in M_n(\mathbb{R})$ . Prove it.

Note that factorizable vectors  $x \otimes y$  are not the whole  $\mathbb{R}^{mn}$ , but span  $\mathbb{R}^{mn}$ .

**4b5 Exercise.**  $SO(m) \otimes SO(n) \subset SO(mn)$ .

Prove it.

Hint: 
$$\langle (x_1 \otimes y_1)(U \otimes V), (x_2 \otimes y_2)(U \otimes V) \rangle = \langle x_1 \otimes y_1, x_2 \otimes y_2 \rangle$$
.

**4b6 Exercise.** (a) If  $(U_t)_{t\in[0,\infty)}$  is a one-parameter semigroup in  $M_n(\mathbb{R})$ , then  $(U_t\otimes U_t)_{t\in[0,\infty)}$  is a one-parameter semigroup in  $M_{n^2}(\mathbb{R})$ .

(b) If A is the generator of  $(U_t)_{t\in[0,\infty)}$  then

$$(A \otimes \mathbf{1}) + (\mathbf{1} \otimes A)$$

is the generator of  $(U_t \otimes U_t)_t$ .

Prove it. Generalize it to  $U_t^{\otimes k}$ .

Hint. (a): use 4b4(b); (b): 
$$(1 + At + o(t)) \otimes (1 + At + o(t)) = \dots$$

In other words,

(4b7) 
$$\exp(A \otimes \mathbf{1} + \mathbf{1} \otimes A) = \exp(A) \otimes \exp(A).$$

In fact,  $\exp(A \otimes \mathbf{1}) = \exp(A) \otimes \mathbf{1}$  and  $\exp(\mathbf{1} \otimes B) = \mathbf{1} \otimes \exp(B)$ , thus  $\exp(A \otimes \mathbf{1} + \mathbf{1} \otimes B) = \exp(A) \otimes \exp(B)$ . Operators  $A \otimes \mathbf{1}$  form an algebra isomorphic to  $M_n(\mathbb{R})$ ; operators  $\mathbf{1} \otimes B$  form another algebra isomorphic to  $M_n(\mathbb{R})$ ; these are two *commuting* copies of  $M_n(\mathbb{R})$  in  $M_{n^2}(\mathbb{R})$ .

## 4c Tensor powers of Brownian rotations

**4c1 Exercise.** (a) If  $(Y_t)_t$  is a Brownian motion in SO(n) then  $(Y_t^{\otimes k})_t$  is a Brownian motion in  $SO(n^k)$ , for any k = 1, 2, ...

(b) If  $(B_t, Y_t)_t$  is a morphism of the standard Brownian motion  $(B_t)_t$  in  $\mathbb{R}$  to a Brownian motion in SO(n), then  $(B_t, Y_t^{\otimes k})_t$  is a morphism of  $(B_t)_t$  to a Brownian motion in  $SO(n^k)$ .

Prove it.

Hint: use 4b4(b) and 4b5.

In fact, the product of *commuting* Brownian rotations is a Brownian rotation. The same holds for morphisms.

**4c2 Exercise.** If  $(Y_t)_t$  is a Brownian motion in SO(n) then  $(\mathbb{E}Y_t)_t$  is a (continuous) one-parameter semigroup in  $M_n(\mathbb{R})$ .

Prove it.

Hint:  $\mathbb{E}(AB) = (\mathbb{E}A)(\mathbb{E}B)$  for independent random matrices A, B.

We have

(4c3) 
$$\mathbb{E}(Y_t^{\otimes k}) = \exp(A_k t) \quad \text{for } t \in [0, \infty), \ k = 1, 2, \dots$$

where  $A_k \in \mathcal{M}_{n^k}(\mathbb{R})$  is the generator of the semigroup.

**4c4 Example.** The isomorphism  $e^{i\alpha} \mapsto \begin{pmatrix} \cos \alpha & \sin \alpha \\ -\sin \alpha & \cos \alpha \end{pmatrix}$  between  $\mathbb{T}$  and SO(2) (mentioned before 3b8) turns  $e^{iB_t}$  into

$$Y_t = \begin{pmatrix} \cos B_t & \sin B_t \\ -\sin B_t & \cos B_t \end{pmatrix} \in SO(2)$$
.

We have

$$\mathbb{E} Y_t = \begin{pmatrix} e^{-t/2} & 0 \\ 0 & e^{-t/2} \end{pmatrix} = \exp(A_1 t), \qquad A_1 = -\frac{1}{2} \cdot \mathbf{1},$$

since  $\mathbb{E}e^{iB_t}=e^{-t/2}$ . Further,

$$Y_t \otimes Y_t = \begin{pmatrix} \cos^2 B_t & \cos B_t \sin B_t & \sin B_t \cos B_t & \sin^2 B_t \\ -\cos B_t \sin B_t & \cos^2 B_t & -\sin^2 B_t & \sin B_t \cos B_t \\ -\sin B_t \cos B_t & -\sin^2 B_t & \cos^2 B_t & \cos B_t \sin B_t \\ \sin^2 B_t & -\sin B_t \cos B_t & -\cos B_t \sin B_t & \cos^2 B_t \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1 + \cos 2B_t & \sin 2B_t & 1 - \cos 2B_t \\ -\sin 2B_t & 1 + \cos 2B_t & -1 + \cos 2B_t & \sin 2B_t \\ -\sin 2B_t & -1 + \cos 2B_t & 1 + \cos 2B_t & \sin 2B_t \\ 1 - \cos 2B_t & -\sin 2B_t & -\sin 2B_t & 1 + \cos 2B_t \end{pmatrix};$$

$$\mathbb{E}(Y_t \otimes Y_t) = \frac{1}{2} \begin{pmatrix} 1 + e^{-2t} & 0 & 0 & 1 - e^{-2t} \\ 0 & 1 + e^{-2t} & -1 + e^{-2t} & 0 \\ 0 & -1 + e^{-2t} & 1 + e^{-2t} & 0 \\ 1 - e^{-2t} & 0 & 0 & 1 + e^{-2t} \end{pmatrix}$$

(it must be a semigroup). For small t,

$$\mathbb{E}(Y_t \otimes Y_t) = \begin{pmatrix} 1 - t & 0 & 0 & t \\ 0 & 1 - t & -t & 0 \\ 0 & -t & 1 - t & 0 \\ t & 0 & 0 & 1 - t \end{pmatrix} + o(t) = 1 + A_2 t + o(t),$$

$$A_2 = \begin{pmatrix} -1 & 0 & 0 & 1 \\ 0 & -1 & -1 & 0 \\ 0 & -1 & -1 & 0 \\ 1 & 0 & 0 & -1 \end{pmatrix},$$

thus  $\mathbb{E}(Y_t \otimes Y_t) = \exp(A_2 t)$ .

#### 4c5 Exercise. Let

$$Y_t = \text{Texp}\left(i\int_0^t (\sigma \, dB_s + v \, ds)\right)$$

for some  $\sigma, v \in iM_n(\mathbb{R}), \ \sigma^* = \sigma, \ v^* = v$ . Then

$$Y_t \otimes Y_t = \operatorname{Texp}\left(i \int_0^t (\sigma \otimes \mathbf{1} + \mathbf{1} \otimes \sigma) dB_s + (v \otimes \mathbf{1} + \mathbf{1} \otimes v) ds\right).$$

Prove it. Generalize it for  $Y_t^{\otimes k}$ .

Hint. According to 4c1(b) and 3b, it must be  $Y_t \otimes Y_t = \text{Texp}(i \int_0^t (\sigma_2 dB_s + v_2 ds))$  for some  $\sigma_2, v_2$ ; for small t we get  $\exp(i\sigma_2 B_t + iv_2 t) = \exp(i\sigma B_t + iv_2 t) \otimes \exp(i\sigma B_t + iv_2 t) + o(t)$ ; recall (4b7).

In fact,

$$\left(\operatorname{Texp}\left(i\int_0^t dX\right)\right)\left(\operatorname{Texp}\left(i\int_0^t dY\right)\right) = \operatorname{Texp}\left(i\int_0^t d(X+Y)\right)$$

whenever X, Y commute.

#### 4c6 Exercise. Let

$$Y_t = \text{Texp}\left(i \int_0^t (\sigma \, dB_s + v \, ds)\right)$$

for some  $\sigma, v \in iM_n(\mathbb{R}), \ \sigma^* = \sigma, \ v^* = v$ . Then

$$A_{1} = -\frac{1}{2}\sigma^{2} + iv,$$

$$A_{2} = -\frac{1}{2}(\sigma \otimes \mathbf{1} + \mathbf{1} \otimes \sigma)^{2} + i(v \otimes \mathbf{1} + \mathbf{1} \otimes v) =$$

$$= -\frac{1}{2}(\sigma^{2} \otimes \mathbf{1} + \mathbf{1} \otimes \sigma^{2}) - \sigma \otimes \sigma + i(v \otimes \mathbf{1} + \mathbf{1} \otimes v),$$

where  $A_k$  are defined by (4c3).

Prove it.

Hint: first, find  $A_1$  by using the asymptotics of  $Y_t$  for small t; second, apply the formula for  $A_1$  to  $Y_t \otimes Y_t$  using 4c5.

**4c7 Example.** Let  $Y_t$  be as in 4c4, then  $i\sigma = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$  is the image of  $i \in \mathbb{C}$  under the embedding  $z \mapsto \begin{pmatrix} \operatorname{Re} z & \operatorname{Im} z \\ -\operatorname{Im} z & \operatorname{Re} z \end{pmatrix}$  (mentioned before 3b8), and v = 0. Using 4c6,

$$(i\sigma)^{2} = -\mathbf{1}; \qquad A_{1} = -\frac{1}{2} \cdot \mathbf{1};$$

$$\sigma^{2} \otimes \mathbf{1} = \mathbf{1} \otimes \mathbf{1} = \mathbf{1}; \qquad \mathbf{1} \otimes \sigma^{2} = \mathbf{1}; \qquad \sigma \otimes \sigma = \begin{pmatrix} 0 & 0 & 0 & -1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ -1 & 0 & 0 & 0 \end{pmatrix};$$

$$A_{2} = -\mathbf{1} - \sigma \otimes \sigma = \begin{pmatrix} -1 & 0 & 0 & 1 \\ 0 & -1 & -1 & 0 \\ 0 & -1 & -1 & 0 \\ 1 & 0 & 0 & -1 \end{pmatrix};$$

the result conforms to 4c4.

Similarly,

$$A_3 = -\frac{1}{2}(\sigma \otimes \mathbf{1} \otimes \mathbf{1} + \mathbf{1} \otimes \sigma \otimes \mathbf{1} + \mathbf{1} \otimes \mathbf{1} \otimes \sigma)^2 + i(v \otimes \mathbf{1} \otimes \mathbf{1} + \mathbf{1} \otimes v \otimes \mathbf{1} + \mathbf{1} \otimes \mathbf{1} \otimes v)$$

and so on. Having all  $A_k$  we know (in principle) all tensor moments of  $Y_t$ , therefore, the distribution of  $Y_t$  (for any t), according to 4b3.

However, Texp  $(i \int (\sigma dB_s + v ds))$  is only a special case of

(4c8) 
$$Y_t = \text{Texp}\left(i \int_0^t (\sigma_1 dB_1(s) + \dots + \sigma_m dB_m(s) + v ds)\right)$$

(recall 3c). By a straightforward generalization of 4c1(b), 4c5, 4c6 we get

$$Y_t \otimes Y_t = \operatorname{Texp}\left(i \int_0^t \left( (\sigma_1 \otimes \mathbf{1} + \mathbf{1} \otimes \sigma_1) dB_1(s) + \dots + (\sigma_m \otimes \mathbf{1} + \mathbf{1} \otimes \sigma_m) dB_m(s) + \right. \\ \left. + \left( v \otimes \mathbf{1} + \mathbf{1} \otimes v \right) ds \right) \right);$$

(4c9) 
$$A_1 = -\frac{1}{2}(\sigma_1^2 + \dots + \sigma_m^2) + iv;$$

$$A_2 = -\frac{1}{2} \sum_{k=1}^m (\sigma_k \otimes \mathbf{1} + \mathbf{1} \otimes \sigma_k)^2 + i(v \otimes \mathbf{1} + \mathbf{1} \otimes v);$$

$$A_3 = -\frac{1}{2} \sum_{k=1}^m (\sigma_k \otimes \mathbf{1} \otimes \mathbf{1} + \mathbf{1} \otimes \sigma_k \otimes \mathbf{1} + \mathbf{1} \otimes \mathbf{1} \otimes \sigma_k)^2 + i(v \otimes \mathbf{1} \otimes \mathbf{1} + \mathbf{1} \otimes v \otimes \mathbf{1} + \mathbf{1} \otimes \mathbf{1} \otimes v);$$

and so on.

## 4d Not just morphisms

We have a satisfactory theory of morphisms. What about a theory of Brownian rotations? Two questions arise naturally.

**4d1.** Whether every Brownian motion in SO(n) is of the form (4c8), or not?

**4d2.** Given two morphisms of the form (4c8), how to decide, whether they give two *identically distributed* Brownian motions in SO(n), or not?

No doubt that two morphisms can represent the same Brownian rotation (I mean, the same distribution). For example,

$$Y_1(t) = \text{Texp}\left(i \int_0^t (\sigma \, dB_s + v \, ds)\right),$$
  
$$Y_2(t) = \text{Texp}\left(i \int_0^t (-\sigma \, dB_s + v \, ds)\right).$$

These two morphisms  $B \to Y_1$ ,  $B \to Y_2$  are connected by an automorphism of B, that is, an isomorphism (invertible morphism)  $B \to B$ ; namely,  $(B_t, -B_t)_t$ . For the two-dimensional  $(B_1(t), B_2(t))_t$  we may use the automorphism

$$(B_1(t), B_2(t)), (B_1(t)\cos\alpha + B_2(t)\sin\alpha, -B_1(t)\sin\alpha + B_2(t)\cos\alpha)$$

where  $\alpha$  is a parameter; we get a continuum of morphisms

$$Y_{\alpha}(t) = \operatorname{Texp}\left(i \int_{0}^{t} \left( (\sigma_{1} \cos \alpha - \sigma_{2} \sin \alpha) dB_{1}(s) + (\sigma_{1} \sin \alpha + \sigma_{2} \cos \alpha) dB_{2}(s) + v ds \right) \right)$$

such that the distribution of  $Y_{\alpha}$  does not depend on the parameter  $\alpha$ . More generally, every rotation of  $\mathbb{R}^m$  (namely, every element of  $O(\mathbb{R}^m)$ ) gives us an automorphism of the standard Brownian motion  $(B_1(t), \ldots, B_m(t))_t$  in  $\mathbb{R}^m$ .

There exist also morphisms of  $(B_1, B_2)$  to  $B_1$ ; here are two examples:

$$\left( (B_1(t), B_2(t)), B_1(t) \right)_t;$$

$$\left( (B_1(t), B_2(t)), \frac{B_1(t) + B_2(t)}{\sqrt{2}} \right)_t.$$

Accordingly, the Brownian rotations

$$Y_1(t) = \text{Texp}\left(i\int_0^t \left(\frac{\sigma}{\sqrt{2}} dB_1(s) + \frac{\sigma}{\sqrt{2}} dB_2(s) + v ds\right)\right),$$
$$Y_2(t) = \text{Texp}\left(i\int_0^t (\sigma dB(s) + v ds)\right)$$

are identically distributed.

## 4e Uniqueness theorem

Generators  $A_k$  (introduced by (4c3)) depend on the distribution of Y (not on a morphism), and determine uniquely the distribution by 4b3. It is an answer to 4d2: given two morphisms, calculate their  $A_1, A_2, \ldots$  by (4c9) and compare them.

Fortunately, it is enough to compare  $A_1, A_2$  only! They determine uniquely  $A_3, A_4, \ldots$  Indeed, knowing  $A_1 = -\frac{1}{2}(\sigma_1^2 + \cdots + \sigma_m^2) + iv$  we know both v (since  $A_1 - A_1^* = 2iv$ ), and  $\sum \sigma_k^2$ . Knowing also  $A_2$  we know  $\sum (\sigma_k \otimes \mathbf{1} + \mathbf{1} \otimes \sigma_k)^2 = (\sum \sigma_k^2) \otimes \mathbf{1} + \mathbf{1} \otimes (\sum \sigma_k^2) + 2\sum \sigma_k \otimes \sigma_k$ , thus, we know  $\sum \sigma_k \otimes \sigma_k$ . Now, in order to find  $A_3$  we need  $\sum (\sigma_k \otimes \mathbf{1} \otimes \mathbf{1} + \mathbf{1} \otimes \sigma_k \otimes \mathbf{1} + \mathbf{1} \otimes \mathbf{1} \otimes \sigma_k)^2 = (\sum \sigma_k^2) \otimes \mathbf{1} \otimes \mathbf{1} + \mathbf{1} \otimes (\sum \sigma_k^2) \otimes \mathbf{1} + \mathbf{1} \otimes \mathbf{1} \otimes (\sum \sigma_k^2) + 2\sum \sigma_k \otimes \sigma_k \otimes \mathbf{1} + 2\sum \sigma_k \otimes \mathbf{1} \otimes \sigma_k + 2\sum \mathbf{1} \otimes \sigma_k \otimes \sigma_k$ , and it is enough to find the three sums  $\sum \sigma_k \otimes \sigma_k \otimes \mathbf{1}, \sum \sigma_k \otimes \mathbf{1} \otimes \sigma_k$  and  $\sum \mathbf{1} \otimes \sigma_k \otimes \sigma_k$ . No problems with the first and the third sum,

$$\sum \sigma_k \otimes \sigma_k \otimes \mathbf{1} = \left(\sum \sigma_k \otimes \sigma_k\right) \otimes \mathbf{1},$$
$$\sum \mathbf{1} \otimes \sigma_k \otimes \sigma_k = \mathbf{1} \otimes \left(\sum \sigma_k \otimes \sigma_k\right).$$

The second sum  $\sum \sigma_k \otimes \mathbf{1} \otimes \sigma_k$  looks worse, but results from  $\sum \sigma_k \otimes \sigma_k$ , too:

$$\big(\sum \sigma_k \otimes \mathbf{1} \otimes \sigma_k\big)_{\alpha\beta\gamma}^{\delta\varepsilon\zeta} = \sum (\sigma_k)_{\alpha}^{\delta} (\mathbf{1})_{\beta}^{\varepsilon} (\sigma_k)_{\gamma}^{\zeta} = (\mathbf{1})_{\beta}^{\varepsilon} \big(\sum \sigma_k \otimes \sigma_k\big)_{\alpha\gamma}^{\delta\zeta}.$$

We see that  $A_1, A_2$  determine  $A_3$  uniquely. The same holds for  $A_4, A_5, \ldots$  and we get a good answer to 4d2.

**4e1 Theorem.** (a) The distribution of the Brownian motion  $(Y_t)_t$  in SO(n) given by

$$Y_t = \operatorname{Texp}\left(i \int_0^t \left(\sigma_1 dB_1(s) + \dots + \sigma_m dB_m(s) + v ds\right)\right)$$

uniquely determines generators  $A_1 \in M_n(\mathbb{R})$ ,  $A_2 \in M_{n^2}(\mathbb{R})$  of the semigroups  $(\mathbb{E}Y_t)_t$ ,  $(\mathbb{E}(Y_t \otimes Y_t))_t$  and is uniquely determined by  $A_1, A_2$ .

(b) The same for the two matrices

$$iv \in \mathcal{M}_n(\mathbb{R}), \qquad \sum_{k=1}^m \sigma_k \otimes \sigma_k \in \mathcal{M}_{n^2}(\mathbb{R}).$$

# 4f Differential operators on the rotation group

The two matrices mentioned in 4e1(b) have an important meaning in terms of differential operators on SO(n).

A smooth function  $f: \mathbb{R}^n \to \mathbb{R}$  has at 0 its gradient (first differential) vector  $\nabla f(0) \in \mathbb{R}^n$ , namely  $(\nabla f(0))_{\alpha} = \frac{\partial}{\partial x_{\alpha}}\big|_{x=0} f(x)$ , and its matrix of second derivatives  $\nabla^2 f(0) \in \mathcal{M}_n(\mathbb{R})$ , namely  $(\nabla f(0))_{\alpha}^{\beta} = \frac{\partial^2}{\partial x_{\alpha} \partial x_{\beta}}\big|_{x=0} f(x)$ . The situation is somewhat more complicated for a smooth function  $f: SO(n) \to \mathbb{R}$  on the smooth manifold SO(n) of dimension  $n^2$ . Given such f, we define  $\nabla f(\mathbf{1}) \in \mathcal{M}_n(\mathbb{R})$  by

(4f1) 
$$(\nabla f(\mathbf{1}))_{\alpha}^{\beta} = \frac{\partial}{\partial A_{\alpha}^{\beta}} \Big|_{A=0} f(\exp A) ;$$

the matrix  $A \in M_n(\mathbb{R})$  may be treated as  $n^2$  variables, and  $A_{\alpha}^{\beta}$  is one of these variables. Similarly we define  $\nabla^2 f(\mathbf{1}) \in M_{n^2}(\mathbb{R})$  by

(4f2) 
$$(\nabla^2 f(\mathbf{1}))_{\alpha,\beta}^{\gamma,\delta} = \frac{\partial^2}{\partial A_{\alpha}^{\gamma} \partial A_{\beta}^{\delta}} \Big|_{A=0} f(\exp A) .$$

4f3 Exercise.

$$f(\exp A) = f(\mathbf{1}) + \langle \nabla f(\mathbf{1}), A \rangle + \frac{1}{2} \langle \nabla^2 f(\mathbf{1}), A \otimes A \rangle + o(\|A\|^2)$$

for  $||A|| \to 0$ ; here  $\langle \cdot, \cdot \rangle$  means<sup>1</sup>

$$\langle A, B \rangle = \sum_{\alpha, \beta} A_{\alpha}^{\beta} B_{\alpha}^{\beta} \quad \text{for } A, B \in \mathcal{M}_{n}(\mathbb{R}) \,,$$

$$\langle A, B \rangle = \sum_{\alpha, \beta, \gamma, \delta} A_{\alpha, \beta}^{\gamma, \delta} B_{\alpha, \beta}^{\gamma, \delta} \quad \text{for } A, B \in \mathcal{M}_{n^{2}}(\mathbb{R}) \,.$$

Prove it.

The following fact generalizes equalities

$$f(B_t) = f(0) + f'(0)B_t + \frac{1}{2}f''(0)B_t^2 + o(t)$$

$$\mathbb{E}f(B_t) = f(0) + \frac{1}{2}f''(0)t + o(t)$$

from  $f: \mathbb{R} \to \mathbb{R}$  to  $f: SO(n) \to \mathbb{R}$ . A matrix  $D \in M_{n^2}(\mathbb{R})$  defined by

$$(4f4) D = -\sum_{k} \sigma_k \otimes \sigma_k$$

will be very useful.

**4f5 Exercise.** Let  $(Y_t)_t$  be given by (4c8); then

$$f(Y_t) = f(\mathbf{1}) + \sum \langle \nabla f(\mathbf{1}), i\sigma_k \rangle B_k(t) + \langle \nabla f(\mathbf{1}), iv \rangle t - \frac{1}{2} \sum \langle \nabla^2 f(\mathbf{1}), \sigma_k \otimes \sigma_k \rangle B_k^2(t) + o(t),$$

$$\mathbb{E} f(Y_t) = f(\mathbf{1}) + \left( \langle \nabla f(\mathbf{1}), iv \rangle + \frac{1}{2} \langle \nabla^2 f(\mathbf{1}), D \rangle \right) t + o(t)$$

for  $t \to 0$ .

Prove it.

We generalize (4f1), (4f2) as follows: for  $U \in SO(n)$ ,

(4f6) 
$$(\nabla f(U))_{\alpha}^{\beta} = \frac{\partial}{\partial A_{\alpha}^{\beta}} \Big|_{A=0} f(U \exp A) ;$$

$$(\nabla^{2} f(U))_{\alpha,\beta}^{\gamma,\delta} = \frac{\partial^{2}}{\partial A_{\alpha}^{\gamma} \partial A_{\beta}^{\delta}} \Big|_{A=0} f(U \exp A) .$$

<sup>&</sup>lt;sup>1</sup>In other words,  $\langle A, B \rangle = \operatorname{tr}(AB^*)$ .

In other words,

(4f7) 
$$\nabla f(U) = \nabla g(\mathbf{1}), \qquad \nabla^2 f(U) = \nabla^2 g(\mathbf{1}),$$
 where  $g(V) = f(UV)$  for all  $V$ .

We consider the convolution semigroup  $(\mu_t)_t$  corresponding to  $(Y_t)_t$ ; that is,  $Y_t \sim \mu_t$ . We define the convolution of a function and a measure (on SO(n)) by

$$(f * \mu)(U) = \int f(UV) \,\mu(dV),$$

then  $\mathbb{E} f(Y_t) = \int f d\mu_t = (f * \mu_t)(\mathbf{1})$ , and 4f5 becomes

$$(f * \mu_t)(\mathbf{1}) = f(\mathbf{1}) + \left( \langle \nabla f(\mathbf{1}), iv \rangle + \frac{1}{2} \langle \nabla^2 f(\mathbf{1}), D \rangle \right) t + o(t).$$

4f8 Exercise. For any  $U \in SO(n)$ ,

$$(f * \mu_t)(U) = f(U) + \left( \langle \nabla f(U), iv \rangle + \frac{1}{2} \langle \nabla^2 f(U), D \rangle \right) t + o(t)$$

for  $t \to 0$ .

Prove it.

Hint: use (4f7) and apply 4f5 to g.

**4f9 Exercise.** Denote  $f_t = f * \mu_t$ , then

$$f_{t+\Delta t}(U) = f_t(U) + \left( \langle \nabla f_t(U), iv \rangle + \frac{1}{2} \langle \nabla^2 f_t(U), D \rangle \right) \Delta t + o(\Delta t)$$

for  $\Delta t \to 0+$  (and t = const).

Prove it.

Hint: apply 4f8 to  $f_t$ .

The simplest functions are linear functions,

(4f10) 
$$f_B(U) = \langle B, U \rangle$$
 for  $U \in SO(n)$ ,

 $B \in \mathcal{M}_n(\mathbb{R})$  being a parameter. We have

$$(f_B * \mu_t)(U) = \int f_B(UV) \,\mu_t(dV) = \int \langle B, UV \rangle \,\mu_t(dV) =$$

$$= \langle B, U \int V \,\mu_t(dV) \rangle = \langle B, U \exp(A_1 t) \rangle = \langle B \exp(A_1^* t), U \rangle = f_{B \exp(A_1^* t)}(U) \,,$$

that is,

(4f11) 
$$f_B * \mu_t = f_{B \exp(A_1^* t)}.$$

Clearly,  $f_t(U) = (f_B * \mu_t)(U) = \langle B \exp(A_1^*t), U \rangle$  is a smooth function of  $(t, U) \in [0, \infty) \times SO(n)$ ; by 4f9 it satisfies the PDE (partial differential equation)

(4f12) 
$$\frac{\partial}{\partial t} f_t(U) = \langle \nabla f_t(U), iv \rangle + \frac{1}{2} \langle \nabla^2 f_t(U), D \rangle.$$

Quadratic functions are of the form

(4f13) 
$$f_B(U) = \langle B, U \otimes U \rangle \quad \text{for } U \in SO(n);$$

this time,  $B \in M_{n^2}(\mathbb{R})$ .

**4f14 Exercise.** For all  $B \in M_{n^2}(\mathbb{R})$  and  $t \in [0, \infty)$ ,

$$f_B * \mu_t = f_{B \exp(A_2^* t)}.$$

Prove it.

The PDE (4f12) holds for quadratic  $f_0$  as well. Similarly, it holds for all polynomials  $f_0$ . By approximation (in  $C^2(SO(n))$ ) it holds for all  $f_0$  of class  $C^2$ .

### 4g Existence theorem

Here is a positive answer to 4d1.<sup>2</sup>

**4g1 Theorem.** For every Brownian motion  $(Y_t)_t$  in SO(n) there exists a morphism of the standard Brownian motion in  $\mathbb{R}^m$  (for some m) to  $(Y_t)_t$ .

The theorem follows from three lemmas, 4g9, 4g12, 4g15. Similarly to the variance

$$Var(X) = \mathbb{E}(X^2) - (\mathbb{E}X)^2 = \mathbb{E}(X - \mathbb{E}X)^2$$

of a random variable  $X \in L_2(\Omega, \mathbb{R})$ , we may introduce the tensor variance

$$(4g2) \qquad \operatorname{Var}(U) = \mathbb{E}\left(U \otimes U\right) - (\mathbb{E}U) \otimes (\mathbb{E}U) = \mathbb{E}\left(\left(U - \mathbb{E}U\right) \otimes \left(U - \mathbb{E}U\right)\right) \in \operatorname{M}_{n^2}(\mathbb{R})$$

of a random matrix  $U \in L_2(\Omega, M_n(\mathbb{R}))$ . Clearly,

$$(\operatorname{Var} U)_{\alpha,\beta}^{\gamma,\delta} = \operatorname{Cov}(U_{\alpha}^{\gamma}, U_{\beta}^{\delta}); \quad (\operatorname{Var} U)_{\alpha,\alpha}^{\beta,\beta} = \operatorname{Var}(U_{\alpha}^{\beta}).$$

For any Brownian motion  $(Y_t)_t$  in SO(n),

$$Var(Y_t) = \exp(tA_2) - \exp(tA_1) \otimes \exp(tA_1) =$$

$$= (\mathbf{1} + tA_2 + o(t)) - (\mathbf{1} + tA_1 + o(t)) \otimes (\mathbf{1} + tA_1 + o(t)) = t(A_2 - A_1 \otimes \mathbf{1} - \mathbf{1} \otimes A_1) + o(t)$$

 $<sup>^2</sup>$ See also: K. Yosida, On Brownian motion in a homogeneous Riemannian space. Pacific J. Math. 2, 263–270 (1952).

for  $t \to 0$ . Introducing

$$(4g3) D = A_2 - A_1 \otimes \mathbf{1} - \mathbf{1} \otimes A_1$$

we get

(4g4) 
$$\operatorname{Var}(Y_t) = tD + o(t) \quad \text{for } t \to 0.$$

Especially, if  $A_1$ ,  $A_2$  are given by (4c9), then (4g3) conforms with (4f4):

(4g5) 
$$D = -\sum_{k} \sigma_{k} \otimes \sigma_{k} = \sum_{k} i \sigma_{k} \otimes i \sigma_{k}.$$

**4g6 Example.** Similarly to 4c4,  $e^{i\sigma B_t + ivt}$  turns into

$$Y_t = \begin{pmatrix} \cos(\sigma B_t + vt) & \sin(\sigma B_t + vt) \\ -\sin(\sigma B_t + vt) & \cos(\sigma B_t + vt) \end{pmatrix} \in SO(2).$$

We have

$$\mathbb{E}Y_t = e^{-\sigma^2 t/2} \cdot \begin{pmatrix} \cos vt & \sin vt \\ -\sin vt & \cos vt \end{pmatrix}; \qquad A_1 = -\frac{\sigma^2}{2} \cdot \mathbf{1} + v \cdot \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix};$$

$$\mathbb{E}(Y_t \otimes Y_t) = \frac{1}{2} \begin{pmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & -1 & 0 \\ 0 & -1 & 1 & 0 \\ 1 & 0 & 0 & 1 \end{pmatrix} + \frac{1}{2} e^{-2\sigma^2 t} \cos 2vt \cdot \begin{pmatrix} 1 & 0 & 0 & -1 \\ 0 & 1 & 1 & 0 \\ 0 & 1 & 1 & 0 \\ -1 & 0 & 0 & 1 \end{pmatrix} + \frac{1}{2} e^{-2\sigma^2 t} \sin 2vt \cdot \begin{pmatrix} 0 & 1 & 1 & 0 \\ -1 & 0 & 0 & 1 \\ -1 & 0 & 0 & 1 \\ 0 & -1 & -1 & 0 \end{pmatrix};$$

$$A_2 = -\sigma^2 \cdot \begin{pmatrix} 1 & 0 & 0 & -1 \\ 0 & 1 & 1 & 0 \\ 0 & 1 & 1 & 0 \\ -1 & 0 & 0 & 1 \end{pmatrix} + v \cdot \begin{pmatrix} 0 & 1 & 1 & 0 \\ -1 & 0 & 0 & 1 \\ -1 & 0 & 0 & 1 \\ 0 & -1 & -1 & 0 \end{pmatrix};$$

$$A_1 \otimes \mathbf{1} + \mathbf{1} \otimes A_1 = -\sigma^2 \cdot \mathbf{1} + v \cdot \begin{pmatrix} 0 & 1 & 1 & 0 \\ -1 & 0 & 0 & 1 \\ 0 & -1 & -1 & 0 \end{pmatrix};$$

$$D = \sigma^2 \cdot \begin{pmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & -1 & 0 \\ 0 & -1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix} = \sigma^2 \cdot \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} \otimes \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}.$$

**4g7 Exercise.**  $\mathbb{E}((Y_t - \mathbf{1}) \otimes (Y_t - \mathbf{1})) = tD + o(t) \text{ for } t \to 0.$ 

Prove it.

Hint: 
$$\mathbb{E}\left((Y_t - \mathbf{1}) \otimes (Y_t - \mathbf{1})\right) - \operatorname{Var}(Y_t) = (\mathbf{1} - \mathbb{E}Y_t) \otimes (\mathbf{1} - \mathbb{E}Y_t).$$

One more formula for D will be given by (4g13).

**4g8 Exercise.**  $\mathbb{E}|Y_t - \mathbf{1}|^2 = O(t)$  for  $t \to 0$ .

Prove it.

Hint: use 4g7.

Here and henceforth | ... | means not only the absolute value of a real or complex number, but also a norm on  $M_n(\mathbb{R})$ . The choice of a norm influences only constants.

Our first lemma is just a linear algebra (rather than probability).

**4g9 Lemma.** Matrices  $A_1 \in M_n(\mathbb{R}), A_2 \in M_{n^2}(\mathbb{R})$  are of the form (4c9) if and only if they satisfy the following two conditions (where D is defined by (4g3)): (a)  $D_{\alpha,\beta}^{\gamma,\delta} = D_{\beta,\alpha}^{\delta,\gamma}$ ,  $D_{\alpha,\beta}^{\gamma,\delta} = -D_{\gamma,\beta}^{\alpha,\delta}$  for all  $\alpha,\beta,\gamma,\delta$ , and

(a) 
$$D_{\alpha,\beta}^{\gamma,\delta} = D_{\beta,\alpha}^{\delta,\gamma}$$
,  $D_{\alpha,\beta}^{\gamma,\delta} = -D_{\gamma,\beta}^{\alpha,\delta}$  for all  $\alpha,\beta,\gamma,\delta$ , and

$$\sum_{\alpha,\beta,\gamma,\delta} D_{\alpha,\beta}^{\gamma,\delta} Z_{\alpha}^{\gamma} Z_{\beta}^{\delta} \ge 0 \quad \text{for all } Z \in \mathcal{M}_n(\mathbb{R}) ;$$

(b) 
$$\sum_{\gamma} D_{\gamma,\alpha}^{\beta,\gamma} = (A_1)_{\alpha}^{\beta} + (A_1)_{\beta}^{\alpha}$$
 for all  $\alpha, \beta$ .

*Proof.* Let  $A_1, A_2$  be of the form (4c9). By (4g5),  $D = \sum_k i\sigma_k \otimes i\sigma_k$ ; it satisfies  $D_{\alpha,\beta}^{\gamma,\delta} = D_{\beta,\alpha}^{\delta,\gamma}$ and  $D_{\alpha,\beta}^{\gamma,\delta} = -D_{\gamma,\beta}^{\alpha,\delta}$  since  $(i\sigma_k)^* = -i\sigma_k$ . Also,

$$\sum_{\alpha,\beta,\gamma,\delta} D_{\alpha,\beta}^{\gamma,\delta} Z_{\alpha}^{\gamma} Z_{\beta}^{\delta} = \sum_{k} \left( \sum_{\alpha,\gamma} (i\sigma_{k})_{\alpha}^{\gamma} Z_{\alpha}^{\gamma} \right) \left( \sum_{\beta,\delta} (i\sigma_{k})_{\beta}^{\delta} Z_{\beta}^{\delta} \right) = \sum_{k} \left( \sum_{\alpha,\gamma} (i\sigma_{k})_{\alpha}^{\gamma} Z_{\alpha}^{\gamma} \right)^{2} \ge 0,$$

thus (a) holds. Further,<sup>4</sup>

$$\sum_{\gamma} D_{\gamma,\alpha}^{\beta,\gamma} = \sum_{k} \sum_{\gamma} (i\sigma_k)_{\gamma}^{\beta} (i\sigma_k)_{\alpha}^{\gamma} = \sum_{k} (i\sigma_k \cdot i\sigma_k)_{\alpha}^{\beta} = \left(-\sum_{k} \sigma_k^2\right)_{\alpha}^{\beta};$$

but also

$$A_1 + A_1^* = -\sum_k \sigma_k^2 \,,$$

thus (b) holds.

Now assume that (a), (b) hold. A positive quadratic form is a sum of squared linear forms:

$$\sum_{\alpha,\beta,\gamma,\delta} D_{\alpha,\beta}^{\gamma,\delta} Z_{\alpha}^{\gamma} Z_{\beta}^{\delta} = \sum_{k} \left( \sum_{\alpha,\beta} (i\sigma_{k})_{\alpha}^{\beta} Z_{\alpha}^{\beta} \right)^{2} \quad \text{for all } Z \in \mathcal{M}_{n}(\mathbb{R})$$

for some  $i\sigma_k \in M_n(\mathbb{R})$ . That is,  $D = \sum_k (i\sigma_k) \otimes (i\sigma_k) = -\sum_k \sigma_k \otimes \sigma_k$ . If  $Z^* = Z$  then

$$\sum_{\alpha,\beta,\gamma,\delta} D_{\alpha,\beta}^{\gamma,\delta} Z_{\alpha}^{\gamma} Z_{\beta}^{\delta} = -\sum_{\alpha,\beta,\gamma,\delta} D_{\gamma,\beta}^{\alpha,\delta} Z_{\alpha}^{\gamma} Z_{\beta}^{\delta} = -\sum_{\alpha,\beta,\gamma,\delta} D_{\gamma,\beta}^{\alpha,\delta} Z_{\gamma}^{\alpha} Z_{\beta}^{\delta} = -\sum_{\alpha,\beta,\gamma,\delta} D_{\alpha,\beta}^{\gamma,\delta} Z_{\alpha}^{\gamma} Z_{\beta}^{\delta}$$

<sup>&</sup>lt;sup>4</sup>The upper index of a matrix is the row number, the lower index is the column number.

must vanish, thus  $\sum_{\alpha,\beta} (i\sigma_k)_{\alpha}^{\beta} Z_{\alpha}^{\beta}$  must vanish, which means that  $(i\sigma_k)_{\beta}^{\alpha} = -(i\sigma_k)_{\alpha}^{\beta}$ , that is,  $(i\sigma_k)^* = -i\sigma_k$  and  $\sigma_k^* = \sigma_k$ .

We have

$$(A_1 + A_1^*)_{\alpha}^{\beta} = \sum_{\gamma} D_{\gamma,\alpha}^{\beta,\gamma} = -\sum_{k} \sum_{\gamma} (\sigma_k)_{\gamma}^{\beta} (\sigma_k)_{\alpha}^{\gamma} = -\sum_{k} (\sigma_k^2)_{\alpha}^{\beta},$$

thus  $A_1 + A_1^* = -\sum_k \sigma_k^2$ . Introducing  $iv \in M_n(\mathbb{R})$  by

$$iv = \frac{1}{2}(A_1 - A_1^*)$$

we get

$$A_1 = \frac{1}{2}(A_1 + A_1^*) + \frac{1}{2}(A_1 - A_1^*) = -\frac{1}{2}\sum_k \sigma_k^2 + iv.$$

Finally,

$$A_{2} = D + A_{1} \otimes \mathbf{1} + \mathbf{1} \otimes A_{1} =$$

$$= -\sum_{k} \sigma_{k} \otimes \sigma_{k} - \frac{1}{2} \sum_{k} (\sigma_{k}^{2} \otimes \mathbf{1} + \mathbf{1} \otimes \sigma_{k}^{2}) + iv \otimes \mathbf{1} + \mathbf{1} \otimes iv =$$

$$= -\frac{1}{2} \sum_{k} (\sigma_{k} \otimes \mathbf{1} + \mathbf{1} \otimes \sigma_{k})^{2} + i(v \otimes \mathbf{1} + \mathbf{1} \otimes v);$$

we see that  $A_1, A_2$  are of the form (4c9).

**4g10 Exercise.** Let random variables  $M_t \geq 0$ , 0 < t < 1, satisfy (for  $t \to 0$ )

- $||M_t||_{L_{\infty}} = O(1);$
- $\bullet \ \|M_t\|_{L_2} = O(\sqrt{t});$
- $\mathbb{P}(M_t \geq \varepsilon) = o(t)$  for every  $\varepsilon > 0$ .

Then  $\mathbb{E} M_t^3 = o(t)$ .

Prove it.

Hint:  $(\min(M_t, \varepsilon))^3 \le \varepsilon M_t^2$ .

**4g11 Exercise.** Let  $(Y_t)_t$  be a Brownian motion in SO(n). Then  $\mathbb{E}|Y_t - \mathbf{1}|^3 = o(t)$  for  $t \to 0$ . Prove it.

Hint: apply 4g10 to  $M_t = |Y_t - \mathbf{1}|$ , taking into account 4g8 and 1e1; SO(n) is not  $\mathbb{R}$ , but still has an invariant metric.

**4g12 Lemma.** For every Brownian motion  $(Y_t)_t$  in SO(n) the matrices  $A_1, A_2$  defined by (4c3) satisfy conditions 4g9(a,b).

*Proof.* Clearly,  $D_{\alpha,\beta}^{\gamma,\delta} = D_{\beta,\alpha}^{\delta,\gamma}$ . Also,

$$\sum_{\alpha,\beta,\gamma,\delta} D_{\alpha,\beta}^{\gamma,\delta} Z_{\alpha}^{\gamma} Z_{\beta}^{\delta} = \frac{d}{dt} \Big|_{t=0} \sum_{\alpha,\beta,\gamma,\delta} (\mathbb{E} Y_{t} \otimes Y_{t} - (\mathbb{E} Y_{t}) \otimes (\mathbb{E} Y_{t}))_{\alpha,\beta}^{\gamma,\delta} Z_{\alpha}^{\gamma} Z_{\beta}^{\delta} =$$

$$= \frac{d}{dt} \Big|_{t=0} \operatorname{Var} \sum_{\alpha,\beta} (Y_{t})_{\alpha}^{\beta} Z_{\alpha}^{\beta} \geq 0.$$

Every  $U \in SO(n)$  satisfies  $\mathbf{1} = UU^* = (\mathbf{1} + (U - \mathbf{1}))(\mathbf{1} + (U^* - \mathbf{1})) = \mathbf{1} + (U - \mathbf{1}) + (U^* - \mathbf{1}) + O(|U - \mathbf{1}|^2)$ , thus  $\frac{1}{2}(U + U^*) = \mathbf{1} + O(|U - \mathbf{1}|^2),$ 

and

$$U = \frac{1}{2}(U + U^*) + \frac{1}{2}(U - U^*) = \mathbf{1} + \frac{1}{2}(U - U^*) + O(|U - \mathbf{1}|^2).$$

We have

$$(Y_t - \mathbf{1}) \otimes (Y_t - \mathbf{1}) = \left(\frac{1}{2}(Y_t - Y_t^*) + O(|Y_t - \mathbf{1}|^2)\right) \otimes \left(\frac{1}{2}(Y_t - Y_t^*) + O(|Y_t - \mathbf{1}|^2)\right) =$$

$$= \frac{1}{4}(Y_t - Y_t^*) \otimes (Y_t - Y_t^*) + O(|Y_t - \mathbf{1}|^3)$$

(with an absolute constant in O(...)); by 4g11,

$$\mathbb{E}\left((Y_t - \mathbf{1}) \otimes (Y_t - \mathbf{1})\right) = \frac{1}{4} \mathbb{E}\left((Y_t - Y_t^*) \otimes (Y_t - Y_t^*)\right) + o(t).$$

Using 4g7,

(4g13) 
$$D = \frac{d}{dt} \Big|_{t=0} \frac{1}{4} \mathbb{E} \left( (Y_t - Y_t^*) \otimes (Y_t - Y_t^*) \right),$$

which ensures  $D_{\alpha,\beta}^{\gamma,\delta} = -D_{\gamma,\beta}^{\alpha,\delta}$  and finishes the proof of (a). For proving (b) we start with the equality  $\mathbb{E} Y_t Y_t^* = 1$ ;

$$\mathbf{1}_{\alpha}^{\beta} = \mathbb{E} \sum_{\gamma} (Y_t)_{\gamma}^{\beta} (Y_t^*)_{\alpha}^{\gamma} = \sum_{\gamma} (\mathbb{E} Y_t \otimes Y_t)_{\gamma,\gamma}^{\beta,\alpha} =$$

$$= \sum_{\gamma} (\mathbf{1} + tA_2 + o(t))_{\gamma,\gamma}^{\beta,\alpha} = \mathbf{1}_{\alpha}^{\beta} + t \sum_{\gamma} (A_2)_{\gamma,\gamma}^{\beta,\alpha} + o(t),$$

which means that

(4g14) 
$$\sum_{\gamma} (A_2)_{\gamma,\gamma}^{\beta,\alpha} = 0 \quad \text{for all } \alpha, \beta.$$

Therefore

$$\sum_{\gamma} D_{\gamma,\alpha}^{\beta,\gamma} = -\sum_{\gamma} D_{\gamma,\gamma}^{\beta,\alpha} = -\sum_{\gamma} (A_2 - A_1 \otimes \mathbf{1} - \mathbf{1} \otimes A_1)_{\gamma,\gamma}^{\beta,\alpha} =$$

$$= -\sum_{\gamma} (A_2)_{\gamma,\gamma}^{\beta,\alpha} + \sum_{\gamma} (A_1)_{\gamma}^{\beta} \mathbf{1}_{\gamma}^{\alpha} + \sum_{\gamma} \mathbf{1}_{\gamma}^{\beta} (A_1)_{\gamma}^{\alpha} = 0 + (A_1)_{\alpha}^{\beta} + (A_1)_{\beta}^{\alpha}.$$

Our third lemma is stronger than the uniqueness theorem 4e1.

**4g15 Lemma.** A Brownian motion in SO(n) is uniquely determined by generators  $A_1, A_2$  (defined by (4c3)).

*Proof.* Similarly to 4e it is sufficient to prove that the higher tensor moment generators  $A_3, A_4, \ldots$  are uniquely determined by  $A_1, A_2$ . (In fact we will see that the relations found in 4e hold in general.) Denoting for convenience  $Y_t - \mathbf{1} = Z_t$  we have

$$e^{tA_3} = \mathbb{E}(Y_t \otimes Y_t \otimes Y_t) = \mathbb{E}((\mathbf{1} + Z_t) \otimes (\mathbf{1} + Z_t)) =$$

$$= \mathbf{1} + \mathbb{E}(Z_t \otimes \mathbf{1} \otimes \mathbf{1} + \text{two such terms}) + \mathbb{E}(Z_t \otimes Z_t \otimes \mathbf{1} + \text{two such terms}) + \mathbb{E}(Z_t \otimes Z_t \otimes Z_t) =$$

$$= \mathbf{1} + (e^{tA_1} - \mathbf{1}) \otimes \mathbf{1} \otimes \mathbf{1} + \text{two such terms} +$$

$$+ (e^{tA_2} - e^{tA_1} \otimes \mathbf{1} - \mathbf{1} \otimes e^{tA_1} + \mathbf{1}) \otimes \mathbf{1} + \text{two such terms} + o(t)$$

by 4g11. That is,

$$\mathbf{1} + tA_3 + o(t) = \mathbf{1} + t(A_1 \otimes \mathbf{1} \otimes \mathbf{1} + \text{two such terms}) + \\ + t(\underbrace{(A_2 - A_1 \otimes \mathbf{1} - \mathbf{1} \otimes A_1)}_{=D} \otimes \mathbf{1} + \text{two such terms}) + o(t);$$

$$A_3 = (A_1 \otimes \mathbf{1} \otimes \mathbf{1} + \text{two such terms}) + (D \otimes \mathbf{1} + \text{two such terms});$$

namely,

$$(A_3)_{\alpha,\beta,\gamma}^{\delta,\varepsilon,\zeta} = (A_1)_{\alpha}^{\delta} \mathbf{1}_{\beta}^{\varepsilon} \mathbf{1}_{\gamma}^{\zeta} + \mathbf{1}_{\alpha}^{\delta} (A_1)_{\beta}^{\varepsilon} \mathbf{1}_{\gamma}^{\zeta} + \mathbf{1}_{\alpha}^{\delta} \mathbf{1}_{\beta}^{\varepsilon} (A_1)_{\gamma}^{\zeta} + D_{\alpha,\beta}^{\delta,\varepsilon} \mathbf{1}_{\gamma}^{\zeta} + D_{\alpha,\gamma}^{\delta,\zeta} \mathbf{1}_{\beta}^{\varepsilon} + D_{\beta,\gamma}^{\varepsilon,\zeta} \mathbf{1}_{\alpha}^{\delta}.$$

The same argument works for  $A_4, A_5, \ldots$