

Gaussian Random Fields

DRAFT

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1 Review

Definition 1.1. Let (Ω, \mathcal{F}, P) be a probability space. Random variables $\{X_1, \dots, X_n\}$ are called *jointly Gaussian* with variance matrix

$$D = [D_{ij}] > 0$$

if

$$d\mu_{X_1 \dots X_n} = (2\pi)^{-n/2} [\det D]^{-1/2} e^{-1/2 \langle x, D^{-1}x \rangle} dx$$

where $x = (x_1, \dots, x_n) \in \mathbb{R}^n$, $dx = dx_1 \cdots dx_n$

Remark 1.2. 1. $\{X_1, \dots, X_n\}$ are Gaussian with variance $D = [D_{ij}]$ if and only if

$$C_{X_1 \dots X_n} = \mathbb{E}(e^{it_1 X_1 + \dots + it_n X_n}) = e^{-1/2 \langle t, Dt \rangle}$$

$$\langle t, Dt \rangle = \sum_{i,j=1}^n t_i t_j D_{ij}, \quad t = (t_1, \dots, t_n)$$

2. $D_{ij} = \mathbb{E}(X_i X_j)$, $\mathbb{E}(X_i) = 0$

3. $\{X_1, \dots, X_n\}$ are Gaussian if and only if

$$\forall \alpha_i \in \mathbb{R}, \quad \alpha_1 X_1 + \dots + \alpha_n X_n \text{ is Gaussian.}$$

4. Let $\{X_1, \dots, X_n\}$ be Gaussian with variance $I_{n \times n}$.

$$\mathbb{E}(X_1^{m_1} X_2^{m_2} \cdots X_n^{m_n}) = \begin{cases} 0 & \text{if } l \text{ is odd} \\ \frac{(2l)!}{2^{(l)}(l)!} & \text{if } l = 2l' \end{cases}$$

where $l = m_1 + \dots + m_n$

2 Gaussian Random Fields

Definition 2.1. Let G be a countable set. The family of random variables $\{X_n\}_{n \in G}$ is called a *Gaussian Random Field (GRF)*, if for any finite subset $\{n_1, \dots, n_k\} \subset G$, the random variables

$$\{X_{n_1}, \dots, X_{n_k}\}$$

are jointly Gaussian.

Remark 2.2. 1. G could be finite, or G could be a singleton, in which case we have a single random variable. However, we only care about the case when G is infinite.

2. Same definition applies for an arbitrary G (e.g. $G = [0, \infty)$ for Brownian motion, say). The only real difference is that some measure theoretic aspects are more delicate.

2.1 Uniqueness

Given a GRF $\{X_n\}_{n \in G}$, we have

$$D_{nm} = \mathbb{E}(X_n X_m), \text{ and } D : G \times G \rightarrow \mathbb{R}, D(n, m) = D_{nm}$$

Theorem 2.3. *Let (Ω, \mathcal{F}, P) and $(\Omega', \mathcal{F}', P')$ be two probability spaces, and $\{X_n\}_{n \in G}$ and $\{X'_n\}_{n \in G'}$ be two GRF on these spaces with variances D_{nm} and D'_{nm} .*

Suppose $D_{nm} \equiv D'_{nm}$, and assume also, that $\mathcal{F}, \mathcal{F}'$ are minimal σ -fields with respect to which $\{X_n\}$ and $\{X'_n\}$ are measurable.

Then, (Ω, \mathcal{F}, P) and $(\Omega', \mathcal{F}', P')$ are isomorphic, and under this isomorphism, X_j corresponds to X'_j .

Remark 2.4. Equivalent formulation:

$$\exists W, \text{ a unitary map, } W : L^2(\Omega, dP) \rightarrow L^2(\Omega', dP')$$

such that

1. W, W^{-1} map bounded functions to bounded functions.
2. $W(XY) = W(X) \cdot W(Y)$ for bounded X, Y .
3. $W(X_n) = X'_n$

2.2 Existence and Basic Model

Let $D : G \times G \rightarrow \mathbb{R}$ be a map, $D(n, m) = D_{nm}$, such that

$$\forall n_1, \dots, n_k \in G, [D_{n_i n_j}]_{1 \leq i, j \leq k} \text{ is strictly positive definite.}$$

Set

$$\Omega = \mathbb{R}^G = \prod_{n \in G} \mathbb{R}, \quad \omega \in \Omega, \{\omega(n)\}_{n \in G}, \omega(n) \in \mathbb{R}$$

and let \mathcal{F} be the Borel σ -algebra generated by cylinders, i.e. sets of the form

$$\{\omega : \omega(n_1) \in B_1, \dots, \omega(n_k) \in B_k\}, \text{ where } B_i \text{ are Borel in } \mathbb{R}$$

If $a_n > 0, \sum_{n \in G} a_n < \infty$ and

$$d(\omega, \omega') = \sum_{n \in G} a_n \frac{|\omega(n) - \omega'(n)|}{1 + |\omega(n) - \omega'(n)|}$$

then d is a metric on Ω , and (Ω, d) is a complete and separable metric space. Moreover, the Borel σ -field generated by the sets open with respect to d is precisely \mathcal{F} .

Given a cylinder $C = \{\omega : \omega(n_1) \in B_1, \dots, \omega(n_k) \in B_k\}$, set

$$P(C) = \mu_{n_1, \dots, n_k}(C) = (2\pi)^{-k/2} (\det D_c)^{1/2} \int_{\prod_{i=1}^k B_i} e^{-\frac{1}{2} \langle x | D_c^{-1} x \rangle} dx$$

where $D_c = [D_{n_i n_j}]_{1 \leq i, j \leq k}$. This a good definition (does not depend on the way we write C) and if we take a field consisting of finite disjoint unions of cylinders, P is a countably additive set function. Hence, by Caratheodory Extension Theorem (or by Kolmogorov Theorem), P extends uniquely to a probability measure on (Ω, \mathcal{F}) .

Thus, (Ω, \mathcal{F}, P) is a probability space such that, by our construction, $X_n(\omega) = \omega(n)$, $n \in G$ is a random variable, and the joint distribution of $\{X_{n_1}, \dots, X_{n_k}\}$ is μ_{n_1, \dots, n_k} , and so, they are Gaussian with variance $D = [D_{ij}]$.

From now on, we will work only with that model:

$$\begin{aligned} \Omega &\rightarrow \mathbb{R}^G \\ \mathcal{F} &\rightarrow \text{Borel } \sigma\text{-field} \\ P &\rightarrow \text{Gaussian measure induced by the marginals } \mu_{n_1, \dots, n_k} \end{aligned}$$

2.3 Support properties of P

Proposition 2.5. *Let $A_n > 0, n \in G$ be a sequence such that*

$$\sum_{n \in G} A_n D_{nn} < \infty$$

and let

$$\Omega' = \{\omega \in \Omega : \sum_{n \in G} A_n \omega^2(n) < \infty\}$$

Then the following holds:

1. Ω' is measurable.

2. $P(\Omega') = 1$

Proof. Property (1) is trivial. To prove (2), let

$$F(\omega) = \sum_{n \in G} A_n \omega^2(n), \quad F : \Omega \rightarrow [0, \infty], \quad F \geq 0$$

Then

$$\begin{aligned} \int_{\Omega} F(\omega) dP(\omega) &= \int_{\Omega} \sum_{n \in G} A_n \omega^2(n) dP(\omega) = \\ &= \sum_{n \in G} A_n \int_{\Omega} \omega^2(n) dP(\omega) = \sum_{n \in G} A_n D_{nn} < \infty \\ &\implies F(\omega) < \infty \quad P\text{-a.e. } \omega \end{aligned}$$

□

2.4 Hilbert Spaces

Consider the Hilbert space

$$l^2(G) = \left\{ \omega : \Omega \rightarrow \mathbb{R} \mid \sum_{n \in G} |\omega(n)|^2 < \infty \right\}$$

equipped with the usual inner product

$$\langle \omega \mid \omega' \rangle = \sum_{n \in G} \omega(n) \omega'(n)$$

Remark 2.6. Usually, $l^2(G)$ denotes the complex version of this Hilbert space. However, in this document, we are mostly considering $l_{\mathbb{R}}^2$, and we decided to simplify our notation by dropping the subscript \mathbb{R} . Hence, to avoid confusion, we will always write $l_{\mathbb{C}}^2$ when referring to the complex space.

The matrix $D = [D_{nm}]$ defines a linear map, which naturally extends to a (possibly unbounded) self-adjoint linear operator on $l^2(G)$. We will assume that both D and D^{-1} are *bounded*. This amounts to say that, if

$$(D\omega)(n) = \sum_{m \in G} D_{nm} \omega(m)$$

then the sum on the right hand side is convergent, and also assuming that for some $M \geq 0$

$$\sum_{n \in G} |(D\omega)(n)|^2 \leq M \sum_{n \in G} |\omega(n)|^2, \quad \forall \omega \in l^2(G)(G)$$

This assures that D is bounded and self-adjoint, or more explicitly, that

$$\begin{cases} \langle \omega | D\omega \rangle \leq \sqrt{M} \|\omega\|^2 \\ \langle \omega' | D\omega \rangle = \langle D\omega' | \omega \rangle \end{cases}$$

To insure the existence and boundedness of D^{-1} , we further need to assume that

$$\exists m > 0, \quad \langle \omega | D\omega \rangle \geq m \|\omega\|^2 \Rightarrow \|D^{-1}\|^2 \leq m$$

In terms of $\{D_{nm}\}$, D is bounded if

$$\sup_{n \in G} \left\{ \sum_{m \in G} |D_{nm}| \right\} < \infty$$

In particular, $\forall n, \quad m \leq D_{nn} \leq \sqrt{M}$, so that

$$\sum_{n \in G} A_n D_{nn} < \infty \iff \sum_{n \in G} A_n < \infty$$

2.5 Explicit Computations with GRF

Let $\alpha \in l^2(G)$, and suppose that only finitely many coordinates are non-zero. The vector space of such α 's is denoted ${}_f l^2_{\mathbb{R}}(G)$.

$${}_f l^2(G) = \{\alpha \in l^2(G) : \alpha(n) \neq 0 \text{ for finitely many } n\}$$

For $\alpha \in {}_f l^2(G)$, set

$$\Phi_{\alpha}(\omega) = \underbrace{\sum_{n \in G} \alpha(n) \omega(n)}_{\text{finite sum!}} = \sum_{n \in G} \alpha(n) X_n(\omega)$$

Hence, $\Phi_{\alpha}(\omega)$ is a Gaussian random variable and since

$$\begin{aligned} \int_{\Omega} \Phi_{\alpha}(\omega)^2 dP(\omega) &= \sum_{n, m \in G} \alpha(n) \alpha(m) \int_{\Omega} \omega(n) \omega(m) dP = \\ &= \sum_{n, m \in G} \alpha(n) \alpha(m) D_{nm} = \langle \alpha | D\alpha \rangle \end{aligned}$$

$\Phi_{\alpha}(\omega)$ has variance $\langle \alpha | D\alpha \rangle$.

Remark 2.7. If $\delta_n(m)$ is the Kronecker delta on $l^2(G)$, i.e.

$$\delta_n(m) = \begin{cases} 1 & \text{if } n = m \\ 0 & \text{if } n \neq m \end{cases}$$

Then obviously, $\delta_n \in {}_f l^2(G)$, and $\Phi_{\delta_n} = X_n$

The map

$${}_f l^2(G) \ni \alpha \mapsto \Phi_\alpha \in L^2(\Omega, dP)$$

is linear (because Φ_α is a *finite* linear combination), and

$$\|\Phi_\alpha\|_{L^2(\Omega, dP)}^2 = \int_{\Omega} \Phi_\alpha(\omega)^2 dP(\omega) = \langle \alpha | D\alpha \rangle \leq \|D\| \cdot \|\alpha\|_{l^2_{\mathbb{R}}(G)}^2$$

This implies that the map $\alpha \mapsto \Phi_\alpha$ is uniformly continuous. Then, by Extension by Uniform continuity Theorem (MATH-354, Analysis 3), it extends uniquely to a bounded linear map

$$l^2(G) \rightarrow L^2(\Omega, dP)$$

and if $\alpha_n \in {}_f l^2(G)$ is such that $\alpha_n \rightarrow \alpha$, then $\Phi_{\alpha_n} \rightarrow \Phi_\alpha$ in $L^2(\Omega, dP)$.

Claim 2.8. For all $\alpha \in l^2(G)$, Φ_α is a Gaussian random variable with variance $\langle \alpha | D\alpha \rangle$.

Proof. We know that $\forall \alpha_n \in {}_f l^2(G)$,

$$\int_{\Omega} e^{it\Phi_{\alpha_n}} dP = e^{-\frac{1}{2}\langle \alpha_n | D\alpha_n \rangle}$$

Furthermore, $\Phi_{\alpha_n} \rightarrow \Phi_\alpha$ in $L^2(\Omega, dP)$ implies that there exists a subsequence $\alpha_{n_k} \rightarrow \alpha$, such that $\Phi_{\alpha_{n_k}} \rightarrow \Phi_\alpha$ P -a.e. ω .

Then, as $\Phi_{\alpha_{n_k}}$ is real, $|e^{it\Phi_{\alpha_{n_k}}}| = 1$, and thus, by Dominated Convergence Theorem,

$$\lim_{k \rightarrow \infty} \int_{\Omega} e^{it\Phi_{\alpha_{n_k}}} dP = \int_{\Omega} e^{it\Phi_\alpha} dP$$

And obviously,

$$\lim_{k \rightarrow \infty} e^{-\frac{1}{2}\langle \alpha_{n_k} | D\alpha_{n_k} \rangle} = e^{-\frac{1}{2}\langle \alpha | D\alpha \rangle}$$

So that we have

$$\int_{\Omega} e^{it\Phi_\alpha} dP = \lim_{k \rightarrow \infty} \int_{\Omega} e^{it\Phi_{\alpha_{n_k}}} dP = \lim_{k \rightarrow \infty} e^{-\frac{1}{2}\langle \alpha_{n_k} | D\alpha_{n_k} \rangle} = e^{-\frac{1}{2}\langle \alpha | D\alpha \rangle}$$

□

Note also that, by the same argument, $\int_{\Omega} \Phi_{\alpha} dP = 0$

Exercise 1. Show that

$$\Phi_{\alpha} = \sum_{n \in G} \langle \delta_n | \alpha \rangle \Phi_{\delta_n}$$

where the sum on the right is converging in L^2 -sense.

Solution. If $\alpha \in {}_f l^2(G)$, then the sum is finite, and the result holds trivially. Now let $\alpha \in l^2(G)$ be fixed and let $\{g_1, g_2, \dots\}$ be a numbering of elements of G . Define

$$\alpha_k(g_i) = \begin{cases} \alpha(g_i) & \text{if } i \leq k \\ 0 & \text{else} \end{cases}$$

Then clearly, $\alpha_k \in {}_f l^2(G)$ and $\alpha_k \rightarrow \alpha$. Hence, $\Phi_{\alpha_k} \rightarrow \Phi_{\alpha}$ in $L^2(\Omega, dP)$ and thus,

$$\sum_{n \in G} \langle \delta_n | \alpha_k \rangle \Phi_{\delta_n} \rightarrow \Phi_{\alpha} \text{ in } L^2(\Omega, dP)$$

Finally, notice that, the k^{th} partial sum of $\sum_{n \in G} \langle \delta_n | \alpha \rangle \Phi_{\delta_n}$ is

$$\sum_{g_1, \dots, g_k} \langle \delta_n | \alpha \rangle \Phi_{\delta_n} = \sum_{n \in G} \langle \delta_n | \alpha_k \rangle \Phi_{\delta_n}$$

Hence, the sequence of partial sums converges in L^2 -sense as required. \square

Proposition 2.9. *If $\{\alpha_1, \dots, \alpha_n\}$ are linearly independent in $l^2(G)$, then $\{\Phi_{\alpha_1}, \dots, \Phi_{\alpha_n}\}$ are jointly Gaussian with variance matrix*

$$[D_{ij}] = [\langle \alpha_i | D \alpha_j \rangle]_{1 \leq i, j \leq n}$$

Proof. The matrix $[\langle \alpha_i | D \alpha_j \rangle]_{1 \leq i, j \leq n}$ is strictly positive definite since

$$\begin{aligned} \sum_{i, j=1}^n \gamma_i \gamma_j \langle \alpha_i | D \alpha_j \rangle &= \left\langle \sum_{i=1}^n \gamma_i \alpha_i \left| \sum_{j=1}^n D \gamma_j \alpha_j \right. \right\rangle = \\ &\left\langle \sum_{i=1}^n \gamma_i \alpha_i \left| D \left(\sum_{j=1}^n \gamma_j \alpha_j \right) \right. \right\rangle \geq \|D\| \left\| \sum_{i=1}^n \gamma_i \alpha_i \right\|^2 > 0 \end{aligned}$$

unless $\sum_i \gamma_i \alpha_i = 0$ (because $\|D\| \geq m > 0$ by assumption). But, by the linear independence of α_i 's, this is possible only if $\gamma_i \equiv 0, \forall i$. Hence

$$\int_{\Omega} e^{i \sum_n t_n \Phi_{\alpha_n}} dP(\omega) = \int_{\Omega} e^{i \Phi_{\sum_n t_n \alpha_n}} dP(\omega) =$$

$$= e^{-\frac{1}{2}\langle \sum_n t_n \alpha_n | D \sum_n t_n \alpha_n \rangle} = e^{-\frac{1}{2} \sum_{i,j} t_i t_j \langle \alpha_i | D \alpha_j \rangle}$$

□

Fact 2.10. For any $z \in \mathbb{C}$,

$$e^{z\Phi_\alpha} \in L^2(\Omega, dP), \text{ and}$$

$$\int_{\Omega} e^{z\Phi_\alpha} dP = \int_{\mathbb{R}} e^{zx} e^{-\frac{1}{2} \frac{x^2}{\langle \alpha | D \alpha \rangle}} dx = e^{-\frac{z^2}{2} \langle \alpha | D \alpha \rangle}$$

3 Trace Class

Background reading Barry and Simon, *Functional Analysis Volume I*, Chapter VI (last section).

3.1 Trace Class Operators

Definition 3.1. Let \mathcal{H} be a real Hilbert space, and let

$$A : \mathcal{H} \rightarrow \mathcal{H}$$

be a bounded self-adjoint operator (as \mathcal{H} is real, self-adjoint simply means symmetric). We say that A is a *trace class operator* if

$$A = \sum_{n=1}^{\infty} \lambda_n \langle f_n | \cdot \rangle f_n$$

where $\lambda_n \in \mathbb{R}$, $\sum_{n=1}^{\infty} |\lambda_n| < \infty$, and $\{f_n\}$ form an orthonormal basis of \mathcal{H}

Note that from this definition it follows that f_n is the eigenvector associated to eigenvalue λ_n , as

$$A f_n = \lambda_n f_n$$

Definition 3.2. Also define the trace of a trace class operator A :

$$\text{Tr}(A) = \sum_{n=1}^{\infty} \lambda_n$$

and the trace norm

$$\|A\|_1 = \sum_{n=1}^{\infty} |\lambda_n|$$

Note that we will continue to use the standard operator norm, and we will denote it $\|\cdot\|$ as usual, while $\|\cdot\|_1$ will now denote the trace norm.

Also, we say that A is *Hilbert-Schmidt* if $\sum_{n=1}^{\infty} |\lambda_n|^2 < \infty$.

↔ For a more detailed introduction to traces and trace norms, see *Trace ideals and its applications* by Barry Simon.

If A is a trace class, the standard operator norm

$$\|A\| = \sup_n |\lambda_n|$$

Note that the supremum is actually achieved, as

$$\sum_{n=1}^{\infty} |\lambda_n| < \infty \Rightarrow |\lambda_n| \rightarrow 0$$

Moreover, $A \geq \inf_n \lambda_n$, meaning that $\forall \psi \in \mathcal{H}$

$$\langle \psi | A \psi \rangle \geq \left(\inf_n \lambda_n \right) \langle \psi | \psi \rangle$$

Now, let A be trace class such that $A > -1$, i.e. $\inf_n \lambda_n > -1$ or equivalently, $\forall n, \lambda_n > -1$. Then,

$$\det(I + A) = \lim_{n \rightarrow \infty} \prod_{k=1}^n (1 + \lambda_k)$$

Claim 3.3. *The limit on the right hand side exists.*

Proof. Write

$$\prod_{k=1}^n (1 + \lambda_k) = e^{\sum_{k=1}^n \ln(1 + \lambda_k)}$$

Now, as

$$\lim_{x \rightarrow 0} \frac{\ln(1 + x)}{x} = 1 \quad (\text{by L'Hospital Rule})$$

For small enough $|x|$, we have that

$$\frac{1}{2}|x| \leq |\ln(1 + x)| \leq 2|x|$$

Now, as $|\lambda_n| \rightarrow 0$, we have that for large enough n ,

$$\frac{1}{2}|\lambda_n| \leq |\ln(1 + \lambda_n)| \leq 2|\lambda_n|$$

and thus,

$$\sum_{n=1}^{\infty} |\lambda_n| < \infty \Rightarrow \sum_{k=1}^{\infty} \ln(1 + \lambda_k) < \infty$$

□

Hence, we have convergence, and we can write

$$\det(I + A) = \prod_{k=1}^{\infty} (1 + \lambda_k)$$

3.2 Variance Induced Inner Products

As before, consider $l^2(G)$, and let D be self-adjoint and positive definite. Define

$$\langle \cdot | \cdot \rangle_D = \langle \cdot | D(\cdot) \rangle = \left\langle D^{\frac{1}{2}}(\cdot) \left| D^{\frac{1}{2}}(\cdot) \right. \right\rangle$$

Then, $\langle \cdot | \cdot \rangle_D$ is an inner product on $l^2(G)$, and $(l^2(G), \langle \cdot | \cdot \rangle_D)$ is a Hilbert space. We will denote it by $l_D^2(G)$.

Note that, if D is bounded the resulting norm is equivalent to the Euclidean norm. Also,

$$l_D^2(G) = l^2(G)$$

Now, let A be a trace class operator on $l_D^2(G)$, and write

$$A = \sum_{n=1}^{\infty} \lambda_n \langle f_n | \cdot \rangle_D f_n$$

where $\{f_n\}$ is an orthonormal basis of $l_D^2(G)$, and $\sum_{n=1}^{\infty} |\lambda_n| < \infty$.

Consider

$$\begin{aligned} D^{\frac{1}{2}} A D^{-\frac{1}{2}} &= \sum_{n=1}^{\infty} \lambda_n \left\langle f_n \left| D^{-\frac{1}{2}}(\cdot) \right. \right\rangle_D D^{\frac{1}{2}} f_n = \\ &= \sum_{n=1}^{\infty} \lambda_n \left\langle f_n \left| D^{\frac{1}{2}}(\cdot) \right. \right\rangle D^{\frac{1}{2}} f_n = \sum_{n=1}^{\infty} \lambda_n \left\langle D^{\frac{1}{2}} f_n \left| \cdot \right. \right\rangle D^{\frac{1}{2}} f_n \end{aligned}$$

Now, the set $\{D^{\frac{1}{2}} f_n\}$ is an orthonormal basis of $l^2(G)$, as

$$\left\langle D^{\frac{1}{2}} f_n \left| D^{\frac{1}{2}} f_m \right. \right\rangle = \langle f_n | f_m \rangle_D = \delta_{nm}$$

And thus, $D^{\frac{1}{2}} A D^{-\frac{1}{2}}$ is trace class on $l^2(G)$ with eigenvalues λ_n and eigenvectors $D^{\frac{1}{2}} f_n$. Conversely, if A is trace class on $l^2(G)$, $D^{-\frac{1}{2}} A D^{\frac{1}{2}}$ is trace class on $l_D^2(G)$ with eigenvectors $D^{-\frac{1}{2}} f_n$.

3.3 Example

Consider $l_D^2(G)$ and let A be a self-adjoint trace class operator such that $A > -1$. Write

$$A = \sum_{n=1}^{\infty} \lambda_n \langle f_n | \cdot \rangle_D f_n$$

where $\{f_n\}$ are orthonormal w.r.t. $\langle \cdot | \cdot \rangle_D$. Now, set

$$F(\omega) = \sum_{n=1}^{\infty} \lambda_n \Phi_{f_n}^2(\omega)$$

Claim 3.4.

$$F \in L^1(\Omega, dP)$$

Proof. By Monotone convergence Theorem, we have

$$\begin{aligned} \int_{\Omega} |F(\omega)| dP(\omega) &\leq \int_{\Omega} \sum_{n=1}^{\infty} |\lambda_n| \Phi_{f_n}^2(\omega) dP(\omega) = \sum_{n=1}^{\infty} |\lambda_n| \int_{\Omega} \Phi_{f_n}^2(\omega) dP(\omega) = \\ &= \sum_{n=1}^{\infty} |\lambda_n| \langle f_n | D f_n \rangle = \sum_{n=1}^{\infty} |\lambda_n| < \infty \end{aligned}$$

□

Example 3.5. Show that

$$\int_{\Omega} e^{-\frac{1}{2}F(\omega)} dP(\omega) = \left[\frac{1}{\det(I + A)} \right]^{1/2}$$

Solution. We wish to compute $\int_{\Omega} e^{-\frac{1}{2}F(\omega)} dP(\omega)$. To this end, let us first compute

$$\int_{\Omega} e^{-\frac{1}{2} \sum_{k=1}^n \lambda_k \Phi_{f_k}^2(\omega)} dP(\omega)$$

By Proposition 2.9, $\{\Phi_{f_k}\}_{k=1}^n$ are jointly Gaussian with variance

$$[\langle f_i | D f_j \rangle]_{1 \leq i, j \leq n} = I_{n \times n}$$

Hence,

$$\int_{\Omega} e^{-\frac{1}{2} \sum_{k=1}^n \lambda_k \Phi_{f_k}^2(\omega)} dP(\omega) = \frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} e^{-\frac{1}{2} \sum_{k=1}^n \lambda_k x_k^2} e^{-\frac{1}{2} \sum_{k=1}^n x_k^2} dx$$

$$\frac{1}{(2\pi)^{n/2}} \int_{\mathbb{R}^n} e^{-\frac{1}{2} \sum_{k=1}^n (\lambda_k+1)x_k^2} dx = \prod_{k=1}^n (1 + \lambda_k)^{-1/2}$$

Taking $n \rightarrow \infty$ in the above, we get that

$$\lim_{n \rightarrow \infty} \int_{\Omega} e^{-\frac{1}{2} \sum_{k=1}^n \lambda_k \Phi_{f_k}^2(\omega)} dP(\omega) = \prod_{k=1}^{\infty} (1 + \lambda_k)^{-1/2} = [\det(I + A)]^{-1/2}$$

Now, since $\lambda_n \rightarrow 0$, there exists n_0 such that $\forall n \geq n_0$, $|\lambda_n| < \frac{1}{2}$. So, set

$$\tilde{F}(\omega) = \sum_{k=1}^{n_0-1} \lambda_k \Phi_{f_k}^2(\omega) - \sum_{k=n_0}^{\infty} |\lambda_k| \Phi_{f_k}^2(\omega)$$

As before, $\tilde{F} \in L^1(\Omega, dP)$, and since $\tilde{F}(\omega) \leq F(\omega)$, we have that

$$\int_{\Omega} e^{-\frac{1}{2}F(\omega)} dP(\omega) \leq \int_{\Omega} e^{-\frac{1}{2}\tilde{F}(\omega)} dP(\omega)$$

As the second sum in $\tilde{F}(\omega)$ is monotone, we can use the Monotone Convergence Theorem and we have

$$\begin{aligned} \int_{\Omega} e^{-\frac{1}{2}\tilde{F}} dP &= \int_{\Omega} \exp\left(-\frac{1}{2} \sum_{k=1}^{n_0-1} \lambda_k \Phi_{f_k}^2(\omega)\right) \exp\left(-\sum_{k=n_0}^{\infty} |\lambda_k| \Phi_{f_k}^2(\omega)\right) dP(\omega) = \\ &\stackrel{\text{MCT}}{=} \lim_{n \rightarrow \infty} \int_{\Omega} \exp\left(-\frac{1}{2} \sum_{k=1}^{n_0-1} \lambda_k \Phi_{f_k}^2(\omega)\right) \exp\left(-\sum_{k=n_0}^n |\lambda_k| \Phi_{f_k}^2(\omega)\right) dP(\omega) = \\ &= \lim_{n \rightarrow \infty} \left[\prod_{k=1}^{n_0-1} (1 + \lambda_k) \right]^{-1/2} \left[\prod_{k=n_0}^n (1 - |\lambda_k|) \right]^{-1/2} \end{aligned}$$

Since, $|\lambda_k| < \frac{1}{2}$, $\forall k \geq n_0$, the second product is non zero, and as $\sum_{n=1}^{\infty} |\lambda_n| < \infty$ and $|\lambda_n| < 1/2$, the limit exists by Claim 3.3. And thus,

$$\int_{\Omega} e^{-\frac{1}{2}\tilde{F}(\omega)} dP(\omega) < \infty \implies \int_{\Omega} e^{-\frac{1}{2}F(\omega)} dP(\omega) < \infty$$

making $e^{-\frac{1}{2}F(\omega)}$ integrable.

Now, as $F(\omega) = \lim_{N \rightarrow \infty} \sum_{n=1}^N \lambda_n \Phi_{f_n}^2(\omega)$, with the sum converging in L^1 -sense, there exists a subsequence $N_k \rightarrow \infty$ such that

$$F(\omega) = \lim_{k \rightarrow \infty} \sum_{n=1}^{N_k} \lambda_n \Phi_{f_n}^2(\omega) \quad P\text{-a.e. } \omega$$

But

$$\forall k \quad e^{-\frac{1}{2} \sum_{n=1}^{N_k} \lambda_n \Phi_{f_n}^2(\omega)} \leq e^{-\frac{1}{2} \tilde{F}} \in L^1$$

So, by Dominated Convergence Theorem,

$$\int_{\Omega} e^{-\frac{1}{2} F(\omega)} dP(\omega) = \lim_{k \rightarrow \infty} \int_{\Omega} e^{-\frac{1}{2} \sum_{n=1}^{N_k} \lambda_n \Phi_{f_n}^2(\omega)} dP(\omega) = \left[\frac{1}{\det(I + A)} \right]^{1/2}$$

□

Case D = I: As seen before, in this case, A is trace class on $l^2(G)$, and so,

$$A = \sum_{n=1}^{\infty} \lambda_n \langle f_n | \cdot \rangle f_n$$

Then, for $\omega \in l^2(G)$, we have that

$$\langle \omega | A \omega \rangle = \sum_{n=1}^{\infty} \lambda_n |\langle f_n | \omega \rangle|^2 = \sum_{n=1}^{\infty} \lambda_n \Phi_{f_n}^2(\omega)$$

This leads us to *define*:

$$\langle \omega | A \omega \rangle \stackrel{\text{def}}{=} \sum_{n=1}^{\infty} \lambda_n \Phi_{f_n}^2(\omega) = F(\omega)$$

Hence, we can rewrite the result of Example 3.5 as

$$\int_{\Omega} e^{-\frac{1}{2} \langle \omega | A \omega \rangle} dP(\omega) = \left[\frac{1}{\det(I + A)} \right]^{1/2}$$

3.4 Variance Induced Measures

Let A be trace class on $l_D^2(G)$ and suppose $A > -1$. As usual, write,

$$A = \sum_{n=1}^{\infty} \lambda_n \langle f_n | \cdot \rangle_D f_n \quad \sum_{n=1}^{\infty} |\lambda_n| < \infty$$

where $\{f_n\}$ is an orthonormal basis of $l_D^2(G)$.

We were prompted to define

$$\langle \omega | A\omega \rangle_D \stackrel{\text{def}}{=} \sum_{n=1}^{\infty} \lambda_n \Phi_{f_n}^2(\omega)$$

And we showed that

1. $\langle \omega | A\omega \rangle_D \in L^1(\Omega, dP)$ (Claim 3.4)
2. $\int_{\Omega} \exp\left(-\frac{1}{2} \langle \omega | A\omega \rangle_D\right) dP(\omega) = [\det(I + A)]^{-1/2}$ (Example 3.5)

We can now introduce the probability measure

$$dQ = [\det(I + A)]^{1/2} e^{-\frac{1}{2} \langle \omega | A\omega \rangle_D} dP$$

and discuss its effect on the GRF.

Claim 3.6. $\{\Phi_{f_{n_1}}, \dots, \Phi_{f_{n_K}}\}$ are jointly Gaussian w.r.t. dQ , with variance matrix

$$[\delta_{ij}(1 + \lambda_{n_i})^{-1}]_{1 \leq i, j \leq K}$$

Proof. Consider

$$\begin{aligned} \mathbb{E}_{dQ} \left[e^{i(t_1 \Phi_{f_{n_1}} + \dots + t_K \Phi_{f_{n_K}})} \right] &= \int_{\Omega} e^{i(t_1 \Phi_{f_{n_1}} + \dots + t_K \Phi_{f_{n_K}})} dQ = \\ &= [\det(I + A)]^{1/2} \int_{\Omega} \exp\left(i \sum_{k=1}^K t_k \Phi_{f_{n_k}}\right) \exp\left(-\frac{1}{2} \sum_{k=1}^{\infty} \lambda_k \Phi_{f_k}^2(\omega)\right) dP = \end{aligned}$$

and $\forall m \in \mathbb{N}$,

$$\left| \exp\left(i \sum_{k=1}^K t_j \Phi_{f_{n_k}}\right) \exp\left(-\frac{1}{2} \sum_{k=1}^m \lambda_k \Phi_{f_k}^2(\omega)\right) \right| = \left| \exp\left(-\frac{1}{2} \sum_{k=1}^m \lambda_k \Phi_{f_k}^2(\omega)\right) \right|$$

Recall that in Exercise 3.5, we have shown that there exists a subsequence $m_j \rightarrow \infty$ such that

$$\sum_{k=1}^{\infty} \lambda_k \Phi_{f_k}^2(\omega) = \lim_{j \rightarrow \infty} \sum_{k=1}^{m_j} \lambda_k \Phi_{f_k}^2(\omega) \quad P\text{-a.e. } \omega$$

$$\text{and } \forall j, \quad \exp\left(-\frac{1}{2} \sum_{k=1}^{m_j} \lambda_k \Phi_{f_k}^2(\omega)\right) \leq \exp\left(-\frac{1}{2} \tilde{F}(\omega)\right) \in L^1$$

Then obviously,

$$\begin{aligned} & \exp\left(i\sum_{k=1}^K t_k \Phi_{f_{n_k}}\right) \exp\left(-\frac{1}{2}\sum_{k=1}^{\infty} \lambda_k \Phi_{f_k}^2(\omega)\right) = \\ & = \lim_{j \rightarrow \infty} \left[\exp\left(i\sum_{k=1}^K t_k \Phi_{f_{n_k}}\right) \exp\left(-\frac{1}{2}\sum_{k=1}^{m_j} \lambda_k \Phi_{f_k}^2(\omega)\right) \right] \end{aligned}$$

Thus, by Dominated Convergence Theorem, we get that

$$\begin{aligned} & \int_{\Omega} \exp\left(i\left(t_1 \Phi_{f_{n_1}} + \dots + t_K \Phi_{f_{n_K}}\right)\right) dQ = \\ & = \lim_{j \rightarrow \infty} \left[[\det(I + A)]^{1/2} \int_{\Omega} \exp\left(i\sum_{k=1}^K t_k \Phi_{f_{n_k}} - \frac{1}{2}\sum_{k=1}^{m_j} \lambda_k \Phi_{f_k}^2(\omega)\right) dP \right] = \\ & = \lim_{j \rightarrow \infty} \left[\frac{\prod_{k=1}^{m_j} (1 + \lambda_k)^{1/2}}{(2\pi)^{m_j/2}} \int_{\mathbb{R}^{m_j}} \exp\left(i\sum_{k=1}^K t_k x_{n_k} - \frac{1}{2}\sum_{k=1}^{m_j} (1 + \lambda_k) x_k^2\right) dx \right] \end{aligned}$$

Now, we integrate out all the coordinates except $\{x_{n_1}, \dots, x_{n_K}\}$, and for all $m_j \geq \max\{n_k : 1 \leq k \leq K\}$ we get

$$\begin{aligned} & \frac{\prod_{k=1}^{m_j} (1 + \lambda_k)^{1/2}}{(2\pi)^{j/2}} \int_{\mathbb{R}^{m_j}} \exp\left(i\sum_{k=1}^K t_k x_{n_k} - \frac{1}{2}\sum_{k=1}^{m_j} (1 + \lambda_k) x_k^2\right) dx = \\ & = \frac{\prod_{k=1}^K (1 + \lambda_{n_k})^{1/2}}{(2\pi)^{K/2}} \int_{\mathbb{R}^K} \exp\left(\sum_{k=1}^K \left[i t_k x_{n_k} - \frac{1 + \lambda_{n_k}}{2} x_{n_k}^2 \right]\right) dx_{n_1} \dots dx_{n_K} \end{aligned}$$

And thus, by uniqueness of characteristic function, $\{\Phi_{f_{n_1}}, \dots, \Phi_{f_{n_K}}\}$ are jointly Gaussian w.r.t. dQ , with variance matrix

$$[\delta_{ij}(1 + \lambda_{n_i})^{-1}]_{1 \leq i, j \leq K}$$

□

Now consider the random variables $\Phi_{\delta_n}(\omega)$ over dQ .

Proposition 3.7. $\{\Phi_{\delta_n}\}_{n=1}^{\infty}$ form a GRF over $(\Omega, \mathcal{F}, dQ)$, with variance $D(I + A)^{-1}$.

Proof. First of all, notice that $\{\Phi_{\delta_n}\}_{n=1}^\infty$ form a GRF over $(\Omega, \mathcal{F}, dP)$ with variance D , essentially by construction. It is obvious, when we apply Proposition 2.9 to any finite subcollection of δ_n 's.

Now, over dQ , consider

$$t_1\Phi_{\delta_{n_1}} + \dots + t_K\Phi_{\delta_{n_K}}$$

Expanding in f_n , we have that $\delta_{n_k} = \sum_{j=1}^\infty \langle f_j | \delta_{n_k} \rangle_D f_j$ and thus,

$$\sum_{k=1}^K t_k \delta_{n_k} = \sum_{j=1}^\infty \left\langle f_j \left| \sum_{k=1}^K t_k \delta_{n_k} \right. \right\rangle_D f_j = \sum_{j=1}^\infty \langle f_j | \delta_t \rangle_D f_j$$

where $\delta_t = \sum_{k=1}^K t_k \delta_{n_k}$. And so,

$$\sum_{k=1}^K t_k \Phi_{\delta_{n_k}} = \Phi_{\sum_{k=1}^K t_k \delta_{n_k}} = \Phi_{\delta_t} = \sum_{j=1}^\infty \langle f_j | \delta_t \rangle_D \Phi_{f_j}$$

with the sum on the right hand side converging in $L^2(\Omega, dQ)$. Therefore, there exists a subsequence $J_l \rightarrow \infty$ such that

$$\Phi_{\delta_t} = \lim_{l \rightarrow \infty} \sum_{j=1}^{J_l} \langle f_j | \delta_t \rangle_D \Phi_{f_j} \quad Q\text{-a.e.}$$

Now, consider

$$\mathbb{E}_{dQ} \left[e^{i(t_1\Phi_{\delta_{n_1}} + \dots + t_K\Phi_{\delta_{n_K}})} \right] = \int_{\Omega} e^{i(t_1\Phi_{\delta_{n_1}} + \dots + t_K\Phi_{\delta_{n_K}})} dQ = \int_{\Omega} e^{i\Phi_{\delta_t}} dQ$$

By Dominated convergence Theorem, (bounded by 1), we get

$$\int_{\Omega} e^{i\Phi_{\delta_t}} dQ = \lim_{l \rightarrow \infty} \int_{\Omega} e^{i\sum_{j=1}^{J_l} \langle f_j | \delta_t \rangle_D \Phi_{f_j}} dQ$$

But by previous claim, Φ_{f_j} 's have variance $[\delta_{ij}(1 + \lambda_{n_i})^{-1}]_{1 \leq i, j \leq k}$ w.r.t. dQ ,

so

$$\begin{aligned}
& \lim_{l \rightarrow \infty} \int_{\Omega} e^{i \sum_{j=1}^{J_l} \langle f_j | \delta_t \rangle_D} \Phi_{f_j} dQ = \\
& = \lim_{l \rightarrow \infty} \exp \left(-\frac{1}{2} \sum_{j=1}^{J_l} [\langle f_j | \delta_t \rangle_D]^2 (1 + \lambda_j)^{-1} \right) = \\
& = \lim_{l \rightarrow \infty} \exp \left(-\frac{1}{2} \sum_{j=1}^{J_l} \left[\langle (I + A)^{-1/2} f_j | \delta_t \rangle_D \right]^2 \right) = \\
& = \lim_{l \rightarrow \infty} \exp \left(-\frac{1}{2} \sum_{j=1}^{J_l} \left[\langle f_j | (I + A)^{-1/2} \delta_t \rangle_D \right]^2 \right) = \\
& = \exp \left(-\frac{1}{2} \sum_{j=1}^{\infty} \left[\langle f_j | (I + A)^{-1/2} \delta_t \rangle_D \right]^2 \right) = \\
& = \exp \left(-\frac{1}{2} \left\| (I + A)^{-1/2} \delta_t \right\|_D^2 \right) = \\
& = \exp \left(-\frac{1}{2} \langle (I + A)^{-1/2} \delta_t | (I + A)^{-1/2} \delta_t \rangle_D \right) = \\
& = \exp \left(-\frac{1}{2} \langle \delta_t | (I + A)^{-1} \delta_t \rangle_D \right) = \\
& = \exp \left(-\frac{1}{2} \langle \delta_t | D(I + A)^{-1} \delta_t \rangle \right) = \\
& = \exp \left(-\frac{1}{2} \sum_{i,j} t_i t_j \langle \delta_{n_i} | D(I + A)^{-1} \delta_{n_j} \rangle \right) =
\end{aligned}$$

And thus, $\{\Phi_{\delta_n}\}_{n=1}^{\infty}$ form a GRF w.r.t. dQ , with variance $D(I + A)^{-1}$. \square

3.5 Mutual Absolute Continuity

Let D_1, D_2 be two variances under usual assumptions, and let P_1, P_2 be the corresponding Gaussian measures on (Ω, \mathcal{F}) . That is $\{\Phi_{\delta_n}\}$ are jointly Gaussian w.r.t. P_i with variance $D_i, i = 1, 2$.

Question: When are P_1 and P_2 mutually absolutely continuous?

Theorem 3.8 (Shale Theorem). *P_1 and P_2 are mutually absolutely continuous if and only if the operator $(D_1^{-1}D_2 - I)$ is trace class on $l_{D_1}^2$.*

Remark 3.9. This is a fundamental result

Remark 3.10.

$$\begin{aligned} \langle \omega | (D_1^{-1}D_2 - I)\omega \rangle_{D_1} &= \langle \omega | D_1(D_1^{-1}D_2 - I)\omega \rangle = \langle \omega | (D_2 - D_1)\omega \rangle \stackrel{\text{sym}}{=} \\ &= \langle (D_2 - D_1)\omega | \omega \rangle = \langle D_1(D_1^{-1}D_2 - I)\omega | \omega \rangle = \langle (D_1^{-1}D_2 - I)\omega | \omega \rangle_{D_1} \end{aligned}$$

That is $(D_1^{-1}D_2 - I)$ is self-adjoint on $l_{D_1}^2$, and the statement of the theorem makes sense.

Remark 3.11. Trace class assumption means:

$$\begin{aligned} D_1^{-1}D_2 - I &= \sum_{n \in G}^{\infty} \lambda_n \langle f_n | \cdot \rangle_{D_1} f_n \\ D_1^{-1/2}D_2 - D_1^{1/2} &= \sum_{n \in G}^{\infty} \lambda_n \langle f_n | \cdot \rangle_{D_1} D_1^{1/2} f_n \\ [D_1^{-1/2}D_2 - D_1^{1/2}] D_1^{-1/2} &= \sum_{n \in G}^{\infty} \lambda_n \langle D_1^{1/2} f_n | \cdot \rangle D_1^{1/2} f_n \end{aligned}$$

As $\{D_1^{1/2}f_n\}$ form an orthonormal basis of $l^2(G)$, $(D_1^{-1/2}D_2D_1^{-1/2} - I)$ is trace class on $l^2(G)$. This implies that $D_1^{-1/2}(D_2 - D_1)D_1^{-1/2}$ is trace class. By properties of trace class operators, this yields that $(D_1 - D_2)$ is trace class.

So, the natural formulation of Shale Theorem is

“ P_1 and P_2 are mutually absolutely continuous if and only if $(D_1 - D_2)$ is trace class on $l^2(G)$.”

Exercise 2. Express the “ $(D_1 - D_2)$ is trace class” condition in terms of matrix elements. In particular, show that

$$\sum_{n,m \in G} \left| [D_1]_{nm} - [D_2]_{nm} \right|^2 < \infty$$

is necessary.

Solution. Suppose D is trace class on $l^2(G)$. WLOG assume $G = \mathbb{N}$ and let $\{f_k\}_{k=1}^{\infty}$ be the eigenbasis and $\{\delta_n\}_{n=1}^{\infty}$ the standard basis. Expanding f_k in terms of the standard basis yields

$$f_k = \sum_n \langle \delta_n | f_k \rangle \delta_n = \sum_n a_n \delta_n$$

Then, by trace class definition, we have

$$\begin{aligned} D &= \sum_k \lambda_k \langle f_k | \cdot \rangle f_k = \sum_k \lambda_k \langle f_k | \cdot \rangle \left(\sum_n a_n \delta_n \right) = \\ &= \sum_{k,n} \lambda_k a_n \left\langle \sum_m a_m \delta_m | \cdot \right\rangle \delta_n = \sum_{k,n,m} \lambda_k a_n a_m \langle \delta_m | \cdot \rangle \delta_n \end{aligned}$$

Hence, formally,

$$\begin{aligned} \sum_{j,l} |D_{jl}|^2 &= \sum_{j,l} |\langle \delta_j | D \delta_l \rangle|^2 = \sum_{j,l} \left| \left\langle \delta_j \left| \sum_{k,n,m} \lambda_k a_n a_m \langle \delta_m | \delta_l \rangle \delta_n \right. \right\rangle \right|^2 = \\ &= \sum_{j,l} |a_l|^2 \left| \left\langle \delta_j \left| \sum_{k,n} \lambda_k a_n \delta_n \right. \right\rangle \right|^2 = \sum_{j,l} |a_l|^2 |a_j|^2 \left| \sum_k \lambda_k \right|^2 = \\ &= \left(\sum_k \lambda_k \right)^2 \sum_{j,l} a_l^2 a_j^2 \end{aligned}$$

Now, by Parseval's Theorem and by the definition of trace class, $\sum_n a_n^2$ and $\sum_k \lambda_k$ are both absolutely convergent. Hence, all the rearrangements of the sums are justified and moreover, the sum on the right hand side is convergent as a Cauchy product of absolutely convergent sums. Thus,

$$\sum_{n,m} |D_{nm}|^2 = \left(\sum_k \lambda_k \right)^2 \sum_{j,l} a_l^2 a_j^2 < \infty$$

Note that this equality only makes sense if D is assumed to be trace class.

Setting $D = D_1 - D_2$ we get the necessary condition for $D_1 - D_2$ to be trace class in terms of matrix coefficients. \square

Proof of Shale Theorem. We will prove only one direction, namely

If $(D_1^{-1} D_2 - I)$ is trace class on $l_{D_1}^2$, then P_1 and P_2 are mutually absolutely continuous.

So, write

$$(D_1^{-1} D_2 - I) = \sum_{n=1}^{\infty} \lambda_n \langle f_n | \cdot \rangle_{D_1} f_n$$

where $\{f_n\}$ form an orthonormal basis of $l_{D_1}^2$, and $\sum_{n=1}^{\infty} |\lambda_n| < \infty$.

Consider now the random variables $\{\Phi_{f_n}\}$ on $L^2(\Omega, dP_1)$.

$$\int_{\Omega} \Phi_{f_n} \Phi_{f_m} dP_1 = \langle f_n | D_1 f_m \rangle = \langle f_n | f_m \rangle_{D_1} = \delta_{nm}$$

Then, as $\{\Phi_{f_n}\}$ are jointly Gaussian w.r.t. dP_1 , we have that

$$\begin{aligned} \mathbb{E}_{dP_1} \left[e^{i \sum_{n=1}^N t_n \Phi_{f_n}} \right] &= \int_{\Omega} e^{i \sum_{n=1}^N t_n \Phi_{f_n}} dP_1 = \\ &= \frac{1}{(2\pi)^{N/2}} \int_{\mathbb{R}^N} \left[e^{i \sum_{n=1}^N t_n x_n} e^{-\frac{1}{2} \sum_{n=1}^N x_n^2} \right] dx_1 \cdots dx_N \end{aligned}$$

Now, $(D_1^{-1} D_2 - I)$ is trace class, so

$$\begin{aligned} \lambda_n \delta_{nm} &= \langle f_n | (D_1^{-1} D_2 - I) f_m \rangle_{D_1} = \langle f_n | D_1 (D_1^{-1} D_2 - I) f_m \rangle = \\ &= \langle f_n | D_2 f_m \rangle - \langle f_n | D_1 f_m \rangle = \langle f_n | D_2 f_m \rangle - \delta_{nm} \\ &\implies \langle f_n | D_2 f_m \rangle = (1 + \lambda_n) \delta_{nm} \end{aligned}$$

So, if we consider $\{\Phi_{f_n}\}$ w.r.t. dP_2 , they have variance $D = [D_{nm}]$, $D_{nm} = (1 + \lambda_n) \delta_{nm}$. Also, notice that $\langle f_n | D f_n \rangle > 0 \implies \lambda_n > -1, \forall n$. We can now compute

$$\begin{aligned} \mathbb{E}_{dP_2} \left[e^{i \sum_{n=1}^N t_n \Phi_{f_n}} \right] &= \int_{\Omega} e^{i \sum_{n=1}^N t_n \Phi_{f_n}} dP_2 = \\ &= \frac{1}{(2\pi)^{N/2}} \left[\frac{1}{\prod_{n=1}^N (1 + \lambda_n)} \right]^{1/2} \int_{\mathbb{R}^N} \left[e^{i \sum_{n=1}^N t_n x_n} e^{-\frac{1}{2} \sum_{n=1}^N \frac{x_n^2}{1 + \lambda_n}} \right] dx_1 \cdots dx_N \end{aligned}$$

Consider

$$\begin{aligned} &\underbrace{\left[\frac{1}{\prod_{k=1}^K (1 + \lambda_{n_k})} \right]^{1/2}}_{=\mathcal{P}_K} \int_{\Omega} e^{i \sum_{k=1}^K t_k \Phi_{f_{n_k}}} e^{-\frac{1}{2} \sum_{k=1}^K \frac{1}{1 + \lambda_{n_k}} \Phi_{f_{n_k}}^2} e^{\frac{1}{2} \sum_{k=1}^K \Phi_{f_{n_k}}^2} dP_1 = \\ &= \frac{\mathcal{P}_K}{(2\pi)^{K/2}} \int_{\mathbb{R}^K} e^{i \sum_{k=1}^K t_k x_{n_k}} e^{-\frac{1}{2} \sum_{k=1}^K \frac{1}{1 + \lambda_{n_k}} x_{n_k}^2} \underbrace{e^{-\frac{1}{2} \sum_{k=1}^K x_{n_k}^2} e^{\frac{1}{2} \sum_{k=1}^K x_{n_k}^2}}_{=1} d\mathbf{x} = \\ &= \frac{\mathcal{P}_K}{(2\pi)^{K/2}} \int_{\mathbb{R}^K} e^{i \sum_{k=1}^K t_k x_{n_k}} e^{-\frac{1}{2} \sum_{k=1}^K \frac{1}{1 + \lambda_{n_k}} x_{n_k}^2} d\mathbf{x} = \int_{\Omega} e^{i \sum_{k=1}^K t_k \Phi_{f_{n_k}}} dP_2 \end{aligned}$$

Thus,

$$\begin{aligned} \mathbb{E}_{dP_2} \left[e^{i \sum_{k=1}^K t_{n_k} \Phi_{f_{n_k}}} \right] &= \int_{\Omega} e^{i \sum_{k=1}^K t_{n_k} \Phi_{f_{n_k}}} dP_2 = \\ &= \lim_{N \rightarrow \infty} \left[\frac{1}{\left[\prod_{n=1}^N (1 + \lambda_n) \right]^{1/2}} \int_{\Omega} e^{i \sum_{k=1}^K t_k \Phi_{f_{n_k}}} e^{\frac{1}{2} \sum_{n=1}^N \frac{\lambda_n}{1 + \lambda_n} \Phi_{f_n}^2} dP_1 \right] \end{aligned}$$

since for $N > \max\{n_k : 1 \leq k \leq K\}$ we simply integrate out all the extra coordinates. Now, by Claim 3.3,

$$\lim_{N \rightarrow \infty} \frac{1}{\left[\prod_{n=1}^N (1 + \lambda_n) \right]^{1/2}} \quad \text{exists since} \quad \begin{cases} \sum_{n=1}^{\infty} |\lambda_n| < \infty \\ \lambda_n > -1 \end{cases}$$

Hence, consider the function

$$F(\omega) = e^{\frac{1}{2} \sum_{n=1}^{\infty} \frac{\lambda_n}{1 + \lambda_n} \Phi_{f_n}^2}$$

We will show that $F \in L^1(\Omega, dP_1)$, so that the limit and the integral can be interchanged by Dominated Convergence Theorem. So, as

$$\int_{\Omega} \left| \sum_{n=1}^{\infty} \frac{\lambda_n}{1 + \lambda_n} \Phi_{f_n}^2 \right| dP_1 \leq \sum_{n=1}^{\infty} \frac{|\lambda_n|}{|1 + \lambda_n|} \underbrace{\int_{\Omega} \Phi_{f_n}^2 dP_1}_{=1} = \sum_{n=1}^{\infty} \frac{|\lambda_n|}{|1 + \lambda_n|} < \infty$$

F is well defined. Hence, set $F_N = e^{\frac{1}{2} \sum_{n=1}^N \frac{\lambda_n}{1 + \lambda_n} \Phi_{f_n}^2}$

As $\sum_{n=1}^{\infty} |\lambda_n| < \infty$, there is $N_0 > 0$ such that $\forall n \geq N_0$, $|\lambda_n| < \frac{1}{3}$. So for $N > N_0$, consider

$$\tilde{F}_N = \exp \left(\frac{1}{2} \sum_{n=1}^{N_0} \frac{\lambda_n}{1 + \lambda_n} \Phi_{f_n}^2 + \frac{1}{2} \sum_{n=N_0+1}^N \frac{|\lambda_n|}{1 + \lambda_n} \Phi_{f_n}^2 \right)$$

Then, \tilde{F}_N is monotonically increasing, and $\int_{\Omega} \tilde{F}_N dP_1 =$

$$\begin{aligned} &= \frac{1}{(2\pi)^{N/2}} \int_{\mathbb{R}^N} e^{\frac{1}{2} \sum_{n=1}^{N_0} \frac{\lambda_n}{1 + \lambda_n} x_n^2} e^{\frac{1}{2} \sum_{n=N_0+1}^N \frac{|\lambda_n|}{1 + \lambda_n} x_n^2} e^{-\frac{1}{2} \sum_{n=1}^N x_n^2} dx = \\ &= \frac{1}{(2\pi)^{N/2}} \int_{\mathbb{R}^N} \exp \left(-\frac{1}{2} \sum_{n=1}^{N_0} \frac{1}{1 + \lambda_n} x_n^2 - \frac{1}{2} \sum_{n=N_0+1}^N \left(1 - \frac{|\lambda_n|}{1 + \lambda_n} \right) x_n^2 \right) dx = \end{aligned}$$

$$= \left[\prod_{n=1}^{N_0} (1 + \lambda_n) \right]^{-1/2} \left[\prod_{n=N_0+1}^N \left(1 - \frac{|\lambda_n|}{1 + \lambda_n} \right) \right]^{-1/2}$$

And

$$\begin{aligned} \lim_{N \rightarrow \infty} \left[\prod_{n=N_0+1}^N \left(1 - \frac{|\lambda_n|}{1 + \lambda_n} \right) \right]^{-1/2} &\leq \lim_{N \rightarrow \infty} \left[\prod_{n=N_0+1}^N \left(1 - \frac{|\lambda_n|}{1 - 1/3} \right) \right]^{-1/2} = \\ &= \lim_{N \rightarrow \infty} \left[\prod_{n=N_0+1}^N \left(1 - \frac{3|\lambda_n|}{2} \right) \right]^{-1/2} < \infty \quad (\text{by Claim 3.3}) \end{aligned}$$

as $\sum_{n=1}^{\infty} \frac{3|\lambda_n|}{2} < \infty$ and $|\lambda_n| < 1/3$. Thus, we have a uniform bound

$$\int_{\Omega} \tilde{F}_N dP_1 \leq \left[\prod_{n=1}^{N_0} (1 + \lambda_n) \right]^{-1/2} \left[\prod_{n=N_0+1}^N \left(1 - \frac{|\lambda_n|}{1 + \lambda_n} \right) \right]^{-1/2} < \infty$$

Hence, as \tilde{F}_N is monotonically increasing, by Monotone Convergence Theorem, we have that:

$$\lim_{N \rightarrow \infty} \int_{\Omega} \tilde{F}_N dP_1 = \int_{\Omega} \left[\lim_{N \rightarrow \infty} \tilde{F}_N \right] dP_1 = \int_{\Omega} \tilde{F} dP_1 < \infty$$

Now, as $F_N \leq \tilde{F}_N \leq \tilde{F} \in L^1(\Omega, dP_1)$ for all $N > N_0$, we apply Dominated Convergence Theorem to get

$$\lim_{N \rightarrow \infty} \int_{\Omega} F_N dP_1 = \int_{\Omega} F dP_1 \leq \int_{\Omega} \tilde{F} dP_1 < \infty \implies F \in L^1(\Omega, dP_1)$$

So that

$$\int_{\Omega} e^{i \sum_{k=1}^K t_k \Phi_{f_{n_k}}} dP_2 = \left[\prod_{n=1}^{\infty} (1 + \lambda_n) \right]^{-1/2} \int_{\Omega} e^{i \sum_{k=1}^K t_k \Phi_{f_{n_k}}} F(\omega) dP_1$$

We will now show that

$$dP_2 = \frac{F(\omega)}{\left[\prod_{n=1}^{\infty} (1 + \lambda_n) \right]^{1/2}} dP_1$$

First, consider $\delta_{n_k} = \sum_{j=1}^{\infty} \langle f_j | \delta_{n_k} \rangle_D f_j$, and

$$\Phi_{\delta_{n_k}} = \sum_{j=1}^{\infty} \langle f_j | \delta_{n_k} \rangle_D \Phi_{f_j}$$

is converging in L^2 -sense in both $L^2(\Omega, dP_1)$ and $L^2(\Omega, dP_2)$, because the expansion is in the same space, namely l_D^2 . Then, by taking two successive subsequences, we get $J_m \rightarrow \infty$ such that

$$\sum_{k=1}^K t_k \Phi_{\delta_{n_k}} = \lim_{m \rightarrow \infty} \sum_{j=1}^{J_m} \left\langle f_j \left| \sum_{k=1}^K t_k \delta_{n_k} \right. \right\rangle_D \Phi_{f_j}$$

both P_1 -a.e. ω and P_2 -a.e. ω , and thus, $\int_{\Omega} \exp\left(i \sum_{k=1}^K t_k \Phi_{\delta_{n_k}}\right) dP_2 =$

$$\begin{aligned} &= \lim_{m \rightarrow \infty} \int_{\Omega} \exp\left[i \sum_{j=1}^{J_m} \left\langle f_j \left| \sum_{k=1}^K t_k \delta_{n_k} \right. \right\rangle_D \Phi_{f_j}\right] dP_2 = \\ &= \lim_{m \rightarrow \infty} \left[\prod_{n=1}^{\infty} (1 + \lambda_n)^{-1/2} \int_{\Omega} \exp\left(i \sum_{j=1}^{J_m} \left\langle f_j \left| \sum_{k=1}^K t_k \delta_{n_k} \right. \right\rangle_D \Phi_{f_j}\right) F dP_1 \right] = \\ &= \left[\prod_{n=1}^{\infty} (1 + \lambda_n) \right]^{-1/2} \int_{\Omega} \exp\left(i \sum_{k=1}^K t_k \Phi_{\delta_{n_k}}\right) F dP_1 \end{aligned}$$

That is the characteristic functions of the marginals of the measures

$$(dP_2) \text{ and } \left(\prod_{n=1}^{\infty} (1 + \lambda_n)^{-1/2} F dP_1 \right)$$

are the same, which implies that the marginals are the same. Thus, these measures coincide on the cylinders, and are therefore the same. \square

Exercise 3. Show that $F \in L^p(\Omega, dP_1)$ for some $p > 1$, and find the optimal p in terms of λ_n 's.

4 Weak Convergence of Gaussian Measures

4.1 General Definition

Let (M, d) be a complete and separable metric space, \mathcal{F} - the Borel σ -algebra on M , and $\{P_t\}_{t>0}$ (or $\{P_n\}_{n \in \mathbb{N}}$) a family of Borel probability measures on (M, \mathcal{F}) .

Definition 4.1. We say that P_t converges weakly to P as $t \rightarrow \infty$ if for any bounded continuous function $F : M \rightarrow \mathbb{R}$,

$$\lim_{t \rightarrow \infty} \int_M F dP_t = \int_M F dP$$

We then write $P_t \Rightarrow P$.

\hookrightarrow For a more detailed discussion of weak convergence, see the first 18 pages of *Convergence of Probability Measures* by Patrick Billingsley.

4.2 The case of GRF

Suppose now, for our setting:

$$\begin{aligned} \Omega &\rightarrow \mathbb{R}^G \\ \mathcal{F} &\rightarrow \text{Borel } \sigma\text{-field} \\ P_t &\rightarrow \text{Gaussian measures with variances } D^{(t)} \\ \Phi_{\delta_n} &\rightarrow \text{GRF over } P_t \text{ with variance } D^{(t)} \quad (\Phi_{\delta_n} = X_n) \end{aligned}$$

For $\alpha \in \ell^2(G)$, the map

$$\text{Phi}_\alpha(\omega) = \sum_{n \in G} \alpha(n) \omega(n)$$

is continuous, and so, $e^{i\Phi_\alpha(\omega)}$ is continuous as well.

So, if $P_t \Rightarrow P$, we must have

$$\int_\Omega e^{i\Phi_\alpha(\omega)} dP_t \longrightarrow \int_\Omega e^{i\Phi_\alpha(\omega)} dP \implies \lim_{t \rightarrow \infty} e^{-1/2 \langle \alpha | D^{(t)} \alpha \rangle} = \int_\Omega e^{i\Phi_\alpha(\omega)} dP$$

We don't want the integral on the right to be neither 0, nor ∞ , so we assume that $\{D^{(t)}\}$ are uniformly bounded, i.e. that there exist $0 < m \leq M$ such that

$$m \leq D^{(t)} \leq M \iff m \|\alpha\|^2 \leq \langle \alpha | D^{(t)} \alpha \rangle \leq M \|\alpha\|^2$$

Then,

$$\lim_{t \rightarrow \infty} \langle \alpha | D^{(t)} \alpha \rangle \text{ exists}$$

By polarization,

$$\lim_{t \rightarrow \infty} \langle \alpha | D^{(t)} \beta \rangle$$

exists for $\alpha, \beta \in \mathcal{L}^2(G)$. By continuity and boundedness, the limit exists $\forall \alpha, \beta \in \mathcal{L}^2(G)$.

If $\alpha = \delta_n$ and $\beta = \delta_m$, $D_{nm} = \lim_{t \rightarrow \infty} D_{nm}^{(t)}$ exists. Let D be an operator with matrix $[D_{nm}]$. Then, D is self-adjoint and bounded, namely

$$m \leq D \leq M$$

Moreover,

$$\forall \alpha, \beta \in \mathcal{L}^2(G) \quad \lim_{t \rightarrow \infty} \langle \alpha | D^{(t)} \beta \rangle = \langle \alpha | D \beta \rangle$$

Hence, D is a good variance, and

$$\lim_{t \rightarrow \infty} e^{-1/2 \langle \alpha | D^{(t)} \alpha \rangle} = e^{-1/2 \langle \alpha | D \alpha \rangle}$$

so that, P is Gaussian with variance D .

Theorem 4.2. *Let P_t be a family of Gaussian measures with variances $D^{(t)}$ such that for some $M, m > 0$,*

$$m \leq D^{(t)} \leq M$$

Then $P_t \Rightarrow P$ if and only if $D^{(t)} \rightarrow D$, and in that case, P variance D .

Proof. We have just shown the \implies direction.

Conversely, if $D^{(t)} \rightarrow D$,

$$\int_{\Omega} e^{i\Phi_{\alpha}(\omega)} dP_t = e^{-1/2 \langle \alpha | D^{(t)} \alpha \rangle} \longrightarrow e^{-1/2 \langle \alpha | D \alpha \rangle} = \int_{\Omega} e^{i\Phi_{\alpha}(\omega)} dP$$

From Billingsley, this implies that for every cylinder C

$$\lim_{t \rightarrow \infty} P_t(C) = P(C)$$

And this implies that $P_t \Rightarrow P$. □

Corollary 4.3. *Let $\{P_t\}_{t>0}$ be a family of Gaussian measures with variances $D^{(t)}$ satisfying,*

$$0 < m \leq D^{(t)} \leq M$$

Then there exists a subsequence $t_n \rightarrow \infty$ such that $P_{t_n} \Rightarrow P$ for a Gaussian P .

Proof. By diagonal argument, extract a subsequence $t_n \rightarrow \infty$ such that

$$D^{(t_n)} \longrightarrow D$$

and by theorem, we are done. □