

Path Integral Quantum Field Theory

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1 Gaussian Integrals

The General Gaussian integral, for a complex, symmetric $n \times n$ matrix A such that $\operatorname{Re} A \geq 0$ and the eigenvalues a_i of A are nonzero, is given by

$$Z_A(b) = \int d^n x \exp \left(-\frac{1}{2} \sum_{i,j=1}^n x_i A_{ij} x_j + \sum_{i=1}^n b_i x_i \right) = (2\pi)^{n/2} (\det A)^{-1/2} \exp \left(\frac{1}{2} \sum_{i,j=1}^n b_i A_{ij}^{-1} b_j \right).$$

(1.1)

Let μ be a measure in \mathbb{R}^n ; we define the expectation value

$$\begin{aligned} \langle F \rangle_\mu &= \int d\mu(x) F(x) \\ &= \int d^n x \Omega(x) F(x). \end{aligned}$$

(1.2)

The measure is normalized so that

$$\int d\mu(x) = 1.$$

(1.3)

We define the generating function

$$Z_\mu(b) = \left\langle e^{(b,x)} \right\rangle_\mu = \int d\mu(x) \exp \left(\sum_{i=1}^n b_i x_i \right),$$

(1.4)

which is a function of the n -dimensional vector b and the measure μ . The integrand can be expanded

$$\exp \left(\sum_{i=1}^n b_i x_i \right) = \sum_{\ell=0}^{\infty} \frac{1}{\ell!} \sum_{i_1 \dots i_\ell=1}^n b_{i_1} \dots b_{i_\ell} x_{i_1} \dots x_{i_\ell}.$$

(1.5)

Therefore, substituting the definition of the correlator

$$\langle x_{i_1} \dots x_{i_\ell} \rangle_\mu = \int d\mu(x) x_{i_1} \dots x_{i_\ell}$$

(1.6)

we obtain

$$Z_\mu(b) = \sum_{\ell=0}^{\infty} \frac{1}{\ell!} \sum_{i_1 \dots i_\ell=1}^n b_{i_1} \dots b_{i_\ell} \langle x_{i_1} \dots x_{i_\ell} \rangle_\mu.$$

(1.7)

Furthermore, it is useful to notice that

$$\frac{\partial}{\partial b_k} Z_\mu(b) = \int d\mu(x) x_k \exp \left(\sum_{i=1}^n b_i x_i \right),$$

(1.8)

which allows the correlators to be written as

$$\langle x_{i_1} \cdots x_{i_\ell} \rangle_\mu = \frac{\partial}{\partial b_{i_1}} \cdots \frac{\partial}{\partial b_{i_\ell}} Z_\mu(b) \Big|_{b=0}. \quad (1.9)$$

Let us consider the Gaussian measure

$$d\mu_0(x) = d^n x \Omega_0(x) = d^n x \mathcal{N}_0 \exp \left(-\frac{1}{2} \sum_{i,j=0}^n x_i A_{ij} x_j \right), \quad (1.10)$$

where the normalization \mathcal{N}_0 is fixed by the normalization of the measure:

$$\mathcal{N}_0 = (2\pi)^{-n/2} (\det A)^{1/2}. \quad (1.11)$$

The generating function in this case can be readily computed using the boxed equation above:

$$Z(b) = \frac{Z_A(b)}{Z_A(0)} = \exp \left(\frac{1}{2} \sum_{i,j=1}^n b_i A_{ij}^{-1} b_j \right). \quad (1.12)$$

In the future, we will drop any subscript A when referring to a Gaussian measure for simplicity. And therefore, defining $\Delta_{ij} = A_{ij}^{-1}$,

$$\langle x_{i_1} \cdots x_{i_\ell} \rangle_0 = \frac{\partial}{\partial b_{i_1}} \cdots \frac{\partial}{\partial b_{i_\ell}} \exp \left(\frac{1}{2} \sum_{i,j=1}^n b_i \Delta_{ij} b_j \right) \Big|_{b=0}.$$

(1.13)

1.1 Wick's Theorem

Let us start with a couple explicit examples, which we will compute in full detail in order to get familiar with the algebraic manipulations.

One-point function: Let k be an integer between 1 and n , we have

$$\begin{aligned} \langle x_k \rangle_0 &= \frac{\partial}{\partial b_k} \exp \left(\frac{1}{2} \sum_{i,j=1}^n b_i \Delta_{ij} b_j \right) \Big|_{b=0} \\ &= \left(\frac{1}{2} \sum_{j=1}^n \Delta_{kj} b_j + \frac{1}{2} \sum_{i=1}^n b_i \Delta_{ik} \right) \exp \left(\frac{1}{2} \sum_{i,j=1}^n b_i \Delta_{ij} b_j \right) \Big|_{b=0} \\ &= \left(\sum_{j=1}^n \Delta_{kj} b_j \right) \exp \left(\frac{1}{2} \sum_{i,j=1}^n b_i \Delta_{ij} b_j \right) \Big|_{b=0} \\ &= 0, \end{aligned} \quad (1.14)$$

where the last equality comes from the fact that the expression is linear in b , and we need to set $b = 0$.

Two-point function: We now consider a pair of indices k, l , and compute

$$\begin{aligned} \langle x_k x_l \rangle_0 &= \frac{\partial}{\partial b_l} \frac{\partial}{\partial b_k} \exp \left(\frac{1}{2} \sum_{i,j=1}^n b_i \Delta_{ij} b_j \right) \Big|_{b=0} \\ &= \left[\Delta_{kl} + \left(\sum_{j=1}^n \Delta_{kj} b_j \right) \left(\sum_{m=1}^n \Delta_{lm} b_m \right) \right] \exp \left(\frac{1}{2} \sum_{i,j=1}^n b_i \Delta_{ij} b_j \right) \Big|_{b=0} \\ &= \Delta_{kl} \end{aligned} \quad (1.15)$$

Thus, the generating function $Z(b)$ provides a systematic way to compute all correlators for a multi-dimensional Gaussian distribution. Having understood the general rule for such a process, we can generate a recipe to compute $\langle x_{i_1} \cdots x_{i_\ell} \rangle_0$, known as *Wick's Theorem*:

- Write down each $x_{i_1} \cdots x_{i_\ell}$ and organize them pairwise (i_p, i_q) . Note that if ℓ must be even for the correlator to be non-zero.
- There are $(\ell - 1) \times (\ell - 3) \times \cdots \times 3 \times 1$ ways of doing this. Sum over all of these possible pairings.
- To each pair (i_p, i_q) associate a factor $\Delta_{i_p i_q}$.

Let us revisit our result for the **two-point function**: for $\langle x_{i_1} x_{i_2} \rangle_0$, there is only one possible pairing (i_1, i_2) . Therefore

$$\langle x_{i_1} x_{i_2} \rangle_0 = \Delta_{i_1 i_2}. \quad (1.16)$$

Four-point function: For $\langle x_{i_1} x_{i_2} x_{i_3} x_{i_4} \rangle_0$, there are three different pairings

$$P = \{ \{ (i_1, i_2), (i_3, i_4) \}, \{ (i_1, i_3), (i_2, i_4) \}, \{ (i_1, i_4), (i_2, i_3) \} \} \quad (1.17)$$

Wick's theorem then yields

$$\langle x_{i_1} x_{i_2} x_{i_3} x_{i_4} \rangle_0 = \Delta_{i_1 i_2} \Delta_{i_3 i_4} + \Delta_{i_1 i_3} \Delta_{i_2 i_4} + \Delta_{i_1 i_4} \Delta_{i_2 i_3}. \quad (1.18)$$

We can also represent these pairings, *Wick contractions*, as should be familiar from canonical quantum field theory:

$$\langle x_i x_j \rangle_0 = \Delta_{ij} = \overline{\square}_{ij} \quad (1.19)$$

1.2 Perturbed Gaussian Measure

Let us now consider a more complicated measure,

$$\Omega(x) = \frac{1}{Z(\lambda)} e^{-S(x, \lambda)}, \quad (1.20)$$

where the normalization is given as usual by

$$Z(\lambda) = \int d^n x e^{-S(x, \lambda)} \quad (1.21)$$

and

$$\begin{aligned} S(x, \lambda) &= \frac{1}{2} \sum_{i,j=1}^n x_i A_{ij} x_j + \lambda V(x) \\ &= S_0(x) + \lambda V(x). \end{aligned} \quad (1.22)$$

We call $V(\lambda)$ the potential term, foreshadowing the physics we will be doing using the perturbed Gaussian measure. Furthermore

$$\begin{aligned} Z(\lambda) &= \sum_{k=0}^{\infty} \frac{(-\lambda)^k}{k!} \int d^n x V^k(x) \exp \left(-\frac{1}{2} \sum_{i,j=1}^n x_i A_{ij} x_j \right) \\ &= Z(0) \sum_{k=0}^{\infty} \frac{(-\lambda)^k}{k!} \langle V^k(x) \rangle_0, \end{aligned} \quad (1.23)$$

where in the first line we have expanded the exponential term containing $e^V(x)$ and in the second line used

$$\langle F(x) \rangle_0 = \frac{1}{Z(0)} \int d^n x \exp \left(-\frac{1}{2} \sum_{i,j=1}^n x_i A_{ij} x_j \right) F(x), \quad (1.24)$$

and $1/Z(0)$ referring to the normalization constant \mathcal{N}_0 from Eq. 1.10. Recalling Eq. 1.13, given that $F(x)$ admits a Taylor expansion

$$\begin{aligned} F(x) &= \sum_{\ell=0}^{\infty} \sum_{i_1 \dots i_{\ell}=1}^n F_{i_1 \dots i_{\ell}} x_{i_1} \dots x_{i_{\ell}} \\ &= F_0 + F_1 x_1 + \dots + F_n x_n \\ &\quad + F_{11} x_1^2 + F_{12} x_1 x_2 + \dots + F_{nn} x_n^2 \\ &\quad + \dots \end{aligned} \tag{1.25}$$

where $F_{i_1 \dots i_{\ell}}$ refer to the constant expansion coefficients, we can define the expectation value of F for a generic measure μ as

$$\langle F(x) \rangle_{\mu} = \sum_{\ell=0}^{\infty} \sum_{i_1 \dots i_{\ell}=1}^n F_{i_1 \dots i_{\ell}} \langle x_{i_1} \dots x_{i_{\ell}} \rangle_{\mu}; \tag{1.26}$$

for the case of a Gaussian measure we obtain

$$\begin{aligned} \langle F(x) \rangle_0 &= \sum_{\ell=0}^{\infty} \sum_{i_1 \dots i_{\ell}=1}^n F_{i_1 \dots i_{\ell}} \frac{\partial}{\partial b_{i_1}} \dots \frac{\partial}{\partial b_{i_{\ell}}} \exp \left(\frac{1}{2} \sum_{i,j=1}^n b_i \Delta_{ij} b_j \right) \Big|_{b=0} \\ &= F \left(\frac{\partial}{\partial b} \right) \exp \left(\frac{1}{2} \sum_{i,j=1}^n b_i \Delta_{ij} b_j \right) \Big|_{b=0}. \end{aligned} \tag{1.27}$$

Thus, returning to our potential term,

$$\begin{aligned} \frac{Z(\lambda)}{Z(0)} &= \left\langle e^{-\lambda V(x)} \right\rangle_0 \\ &= \exp \left[-\lambda V \left(\frac{\partial}{\partial b} \right) \right] \exp \left(\frac{1}{2} \sum_{i,j=1}^n b_i \Delta_{ij} b_j \right) \Big|_{b=0}. \end{aligned} \tag{1.28}$$

1.3 Perturbed Gaussian Correlators

The perturbative treatment discussed above can be extended to compute moments of the distribution:

$$\langle x_{i_1} \dots x_{i_{\ell}} \rangle = \frac{\int d^n x e^{-S(x,\lambda)} x_{i_1} \dots x_{i_{\ell}}}{\int d^n x e^{-S(x,\lambda)}} = \frac{1}{Z(\lambda)} \int d^n x e^{-S(x,\lambda)} x_{i_1} \dots x_{i_{\ell}} \tag{1.29}$$

which is often referred to as an ℓ -point correlator/function. Let us start our discussion with a simple example.

Two-point function: we need to evaluate

$$\begin{aligned} \int d^n x e^{-S(x,\lambda)} x_{i_1} x_{i_2} &= \int d^n x e^{-S_0(x,\lambda)} \left[\sum_{k=0}^n \frac{(-\lambda)^k}{k!} \lambda^k V^k(x) \right] x_{i_1} x_{i_2} \\ &= Z(0) \sum_{k=0}^n \frac{(-\lambda)^k}{k!} \langle V^k(x) x_{i_1} x_{i_2} \rangle, \end{aligned} \tag{1.30}$$

where in the second line we have expressed the initial correlator in terms of correlators computed in the Gaussian theory, denoted by $\langle \dots \rangle_0$. The Gaussian correlators can be computed using Wick's theorem as before. We shall consider again a quartic potential

$$V(x) = \frac{1}{4!} \sum_{i=1}^n x_i^4, \tag{1.31}$$

and compute all the terms in Eq. 1.30 order by order in λ up to order λ^2 .

- $\mathcal{O}(\lambda^0)$: For $k = 0$ we simply get the two-point Gaussian correlator

$$\langle x_{i_1} x_{i_2} \rangle = \Delta_{i_1 i_2}. \quad (1.32)$$

- $\mathcal{O}(\lambda^1)$: At first order in λ we have one insertion of V :

$$\langle V(x) x_{i_1} x_{i_2} \rangle_0 = \frac{1}{4!} \sum_{i=1}^n \langle x_i^4 x_{i_1} x_{i_2} \rangle_0. \quad (1.33)$$

This Gaussian expectation value involving six factors of x can be evaluated using Wick's theorem. There are two types of contractions.

1. x_1 is contracted with x_2 , and the four x_i are contracted amongst themselves:

$$\overbrace{x_{i_1} x_{i_2}}^{\text{contracted}} (\overbrace{x_i x_i x_i x_i}^{\text{contracted}} + \overbrace{x_i x_i x_i x_i}^{\text{contracted}} + \overbrace{x_i x_i x_i x_i}^{\text{contracted}}) = \Delta_{i_1 i_2} \langle x_i^4 \rangle_0 \quad (1.34)$$

2. x_1 and x_2 are contracted with some of the x_i ; there are 12 such contractions:

$$\overbrace{x_{i_1} x_{i_2} x_i x_i x_i x_i}^{\text{contracted}} + \dots = \Delta_{i i_1} \Delta_{i i_2} \Delta_{i i} \times 4 \times 3. \quad (1.35)$$

Collecting all the terms yields

$$\frac{1}{4!} \sum_{i=1}^n \langle x_i^4 x_{i_1} x_{i_2} \rangle_0 = \Delta_{i_1 i_2} \frac{1}{4!} \sum_{i=1}^n \langle x_i^4 \rangle_0 + \frac{1}{4!} \times 4 \times 3 \sum_{i=1}^n \Delta_{i i_1} \Delta_{i i_2} \Delta_{i i}. \quad (1.36)$$

- $\mathcal{O}(\lambda^2)$: At this order we need to evaluate

$$\frac{1}{2!} \langle V(x)^2 x_{i_1} x_{i_2} \rangle_0 = \frac{1}{2!} \frac{1}{(4!)^2} \sum_{i,j=1}^n \langle x_i^4 x_j^4 x_{i_1} x_{i_2} \rangle_0. \quad (1.37)$$

There are five different types of contractions, each of them coming with a given multiplicity. We encourage the interested reader to compute those contributions, and check carefully that the correct multiplicities are recovered. Collecting all contributions yields

$$\begin{aligned} \int d^n x e^{-S(x,\lambda)} x_{i_1} x_{i_2} &= Z(0) \left[\Delta_{i_1 i_2} - \lambda \left(\Delta_{i_1 i_2} \frac{1}{4!} \sum_{i=1}^n \langle x_i^4 \rangle_0 + \frac{1}{2} \sum_{i=1}^n \Delta_{i i_1} \Delta_{i i_2} \Delta_{i i} \right) \right. \\ &\quad + \lambda^2 \left(\frac{1}{2!} \Delta_{i_1 i_2} \frac{1}{(4!)^2} \sum_{i,j=1}^n \langle x_i^4 x_j^4 \rangle_0 + \frac{1}{2!} \sum_{i=1}^n \Delta_{i i_1} \Delta_{i i_2} \Delta_{i i} \frac{1}{4!} \sum_{j=0}^n \langle x_j^4 \rangle_0 \right. \\ &\quad + \frac{1}{4} \sum_{i,j=1}^n \Delta_{i i_1} \Delta_{i i_2} \Delta_{i j}^2 \Delta_{j j} + \frac{1}{6} \sum_{i,j=1}^n \Delta_{i i_1} \Delta_{j i_2} \Delta_{i j}^3 \\ &\quad \left. \left. + \frac{1}{4} \sum_{i,j=1}^n \Delta_{i i_1} \Delta_{j i_2} \Delta_{i j} \Delta_{i i} \Delta_{j j} \right) \right]. \end{aligned} \quad (1.38)$$

A term that factorises as the product of a subdiagram with external lines and a subdiagram that is made of loops only is called a *vacuum contribution*. For instance, $\Delta_{i_1 i_2} \frac{1}{4!} \sum_{i=1}^n \langle x_i^4 \rangle_0$ is a vacuum contribution because $\langle x_i^4 \rangle_0$ is one.

Finally, we need to divide this expression by $Z(\lambda)$ in order to obtain the two-point correlator as defined in Eq. 1.29. As a result, we obtain a factor $Z(0)/Z(\lambda)$ multiplying the expression inside the square bracket in Eq. 1.38. The ratio $Z(0)/Z(\lambda)$ may be computed using Eq. 1.28 and expanding as $1/(1-x) = 1 + x + x^2 + \dots$ and sorting the terms by powers of λ to obtain

$$Z(0)/Z(\lambda) = 1 + \frac{1}{4!} \sum_{i=1}^n \langle x_i^4 \rangle_0 - \frac{1}{2!} \frac{1}{(4!)^2} \lambda^2 \sum_{i,j=1}^n \langle x_i^4 x_j^4 \rangle_0 + \frac{1}{(4!)^2} \lambda^2 \left(\sum_{i=1}^n \langle x_i^4 \rangle_0 \right)^2 + \mathcal{O}(\lambda^3) \quad (1.39)$$

This cancels all vacuum contributions.

1.4 Generating Functions for the Perturbed Gaussian Measure

We can now generalise the idea of a generating function to the case of a non-Gaussian measure. Introducing

$$Z(b, \lambda) = \int d^n x \exp[-S(x, \lambda) + b_i x_i], \quad (1.40)$$

and remembering that we can also write

$$\langle e^{b_i x_i} \rangle = Z(b, \lambda) / Z(\lambda), \quad (1.41)$$

the correlators in the perturbed measure are obtained by differentiation

$$\langle x_{i_1} \cdots x_{i_\ell} \rangle = \frac{1}{Z(\lambda)} \frac{\partial}{\partial b_{i_1}} \cdots \frac{\partial}{\partial b_{i_\ell}} Z(b, \lambda) \Big|_{b=0}. \quad (1.42)$$

The logarithm of $Z(b, \lambda)$ is usually denoted $W(b, \lambda)$,

$$Z(b, \lambda) = e^{W(b, \lambda)}; \quad (1.43)$$

$W(b, \lambda)$ is the generator of the connected ℓ -point correlators $W_{i_1 \dots i_\ell}$, i.e. the correlators that can be represented as a single diagram, with ℓ open ends:

$$W_{i_1 \dots i_\ell} = \frac{\partial}{\partial b_{i_1}} \cdots \frac{\partial}{\partial b_{i_\ell}} W(b, \lambda) \Big|_{b=0}. \quad (1.44)$$

In statistics, the $W_{i_1 \dots i_\ell}$ are called the *cumulants* of the probability distribution $e^{-S(x, \lambda)}$.