

## I.1. The Gaussian Integral

### Basic Form

Let

$$\mathcal{Z}(\mathbf{A}) = \int_{-\infty}^{\infty} dx_1 \dots \int_{-\infty}^{\infty} dx_n e^{-A_2(\mathbf{x})} \equiv \int d^n x e^{-A_2(\mathbf{x})} \quad (1.1)$$

where  $\mathbf{A}$  is a real, symmetric matrix &

$$A_2(\mathbf{x}) = \frac{1}{2} \mathbf{x}^T \mathbf{A} \mathbf{x} = \frac{1}{2} \sum_{i,j=1}^n x_i A_{ij} x_j \quad (1.2)$$

$\mathbf{A}$  is real & symmetric

→  $\mathbf{A}$  is diagonalizable by a orthogonal transformation, i.e., there exists

$$\mathbf{O}^{-1} = \mathbf{O}^T$$

such that

$$\mathbf{O} \mathbf{A} \mathbf{O}^T = \boldsymbol{\lambda} = \text{diag}(\lambda_i)$$

Note that  $\lambda_i$  are all real.

$$\therefore A_2(\mathbf{x}) = \frac{1}{2} \mathbf{x}^T \mathbf{O}^T \boldsymbol{\lambda} \mathbf{O} \mathbf{x} = \frac{1}{2} \mathbf{y}^T \boldsymbol{\lambda} \mathbf{y} = \frac{1}{2} \sum_{i=1}^n \lambda_i y_i^2$$

where

$$\mathbf{y} = \mathbf{O} \mathbf{x}$$

$$d^n x = \frac{\partial (y_1, \dots, y_n)}{\partial (x_1, \dots, x_n)} d^n y = \det | \mathbf{O} | d^n y = d^n y$$

$$\begin{aligned} \rightarrow \mathcal{Z}(\mathbf{A}) &= \int d^n y \exp\left(-\frac{1}{2} \sum_{i=1}^n \lambda_i y_i^2\right) \\ &= \prod_{i=1}^n \int_{-\infty}^{\infty} dy_i \exp\left(-\frac{1}{2} \lambda_i y_i^2\right) \end{aligned}$$

Provided  $\mathbf{A}$  is positive definite, i.e.,

$$\lambda_i > 0 \quad \forall i$$

the integrals are well defined.

Using

$$\int_{-\infty}^{\infty} dy \exp\left(-\frac{1}{2} a y^2\right) = \sqrt{\frac{2\pi}{a}}$$

we have

$$\mathcal{Z}(\mathbf{A}) = \prod_{i=1}^n \sqrt{\frac{2\pi}{\lambda_i}} = (2\pi)^{n/2} \frac{1}{\sqrt{\prod_{i=1}^n \lambda_i}} = \frac{(2\pi)^{n/2}}{\sqrt{\det \mathbf{A}}} \quad (1.3)$$

To summarize, for a real, symmetric & positive definite  $\mathbf{A}$ ,

$$\int d^n x \exp\left(-\frac{1}{2} \mathbf{x}^T \mathbf{A} \mathbf{x}\right) = \frac{(2\pi)^{n/2}}{\sqrt{\det \mathbf{A}}}$$

If  $\mathbf{A}$  is hermitian & positive definite, then it can be diagonalized by a unitary transformation

$$\mathbf{U} \mathbf{A} \mathbf{U}^+ = \boldsymbol{\lambda} = \text{diag}(\lambda_i) \quad \text{with} \quad \lambda_i > 0 \quad \forall i$$

$$\rightarrow \int d^n x \exp\left(-\frac{1}{2} \mathbf{x}^T \mathbf{A} \mathbf{x}\right) = J \frac{(2\pi)^{n/2}}{\sqrt{\det \mathbf{A}}}$$

where  $|J| = |\det \mathbf{U}| = 1$ .

If  $\mathbf{A}$  is normal & positive definite, then it can be diagonalized by a similarity transformation

$$\mathbf{S} \mathbf{A} \mathbf{S}^{-1} = \boldsymbol{\lambda} = \text{diag}(\lambda_i) \quad \text{with} \quad \lambda_i > 0 \quad \forall i$$

$$\rightarrow \int d^n x \exp\left(-\frac{1}{2} \mathbf{x}^T \mathbf{A} \mathbf{x}\right) = J \frac{(2\pi)^{n/2}}{\sqrt{\det \mathbf{A}}}$$

where  $J = \det \mathbf{S}$ .

## General Form

General form of a Gaussian integral is:

$$\mathcal{Z}(\mathbf{A}, \mathbf{b}) = \int d^n x e^{-J(\mathbf{A}, \mathbf{b}; \mathbf{x})} \quad (1.4)$$

where

$$\begin{aligned} J(\mathbf{A}, \mathbf{b}; \mathbf{x}) &= \mathbf{A}_2(\mathbf{x}) - \mathbf{b}^T \mathbf{x} \\ &= \frac{1}{2} \mathbf{x}^T \mathbf{A} \mathbf{x} - \mathbf{b}^T \mathbf{x} \\ &= \frac{1}{2} \sum_{i,j=1}^n x_i A_{ij} x_j - \sum_{i=1}^n b_i x_i \end{aligned}$$

Minima of  $J$  satisfy

$$\begin{aligned} \frac{\partial J}{\partial x_k} &= 0 \quad \forall k \\ &= \frac{1}{2} \left( \sum_{j=1}^n A_{kj} x_j + \sum_{i=1}^n x_i A_{ik} \right) - b_k \\ &= \sum_{j=1}^n A_{kj} x_j - b_k \end{aligned}$$

$$\begin{aligned} \lambda_i > 0 &\rightarrow \det \mathbf{A} \neq 0 \\ &\rightarrow \mathbf{A}^{-1} \text{ exists} \end{aligned}$$

$\therefore$  Minima of  $J$  are given by

$$\begin{aligned} \mathbf{A} \mathbf{x}^0 &= \mathbf{b} \\ \rightarrow \mathbf{x}^0 &= \mathbf{A}^{-1} \mathbf{b} \\ x_i^0 &= \sum_{j=1}^n (\mathbf{A}^{-1})_{ij} b_j \end{aligned} \quad (1.5)$$

Let

$$\mathbf{x} = \mathbf{x}^0 + \mathbf{y} = \mathbf{A}^{-1} \mathbf{b} + \mathbf{y} \quad (1.6)$$

$\rightarrow$

$$d^n x = d^n y$$

Assuming  $\mathbf{A}^T = \mathbf{A}$ ,

$$\begin{aligned}
J &= \frac{1}{2} (\mathbf{b}^T \mathbf{A}^{-1} + \mathbf{y}^T) \mathbf{A} (\mathbf{A}^{-1} \mathbf{b} + \mathbf{y}) - \mathbf{b}^T (\mathbf{A}^{-1} \mathbf{b} + \mathbf{y}) \\
&= \frac{1}{2} (\mathbf{b}^T \mathbf{A}^{-1} \mathbf{b} + \mathbf{b}^T \mathbf{y} + \mathbf{y}^T \mathbf{b} + \mathbf{y}^T \mathbf{A} \mathbf{y}) - \mathbf{b}^T \mathbf{A}^{-1} \mathbf{b} - \mathbf{b}^T \mathbf{y} \\
\mathbf{b}^T \mathbf{y} &= \mathbf{y}^T \mathbf{b} = \sum_{i=1}^n b_i y_i \\
\rightarrow J &= -\frac{1}{2} \mathbf{b}^T \mathbf{A}^{-1} \mathbf{b} + \frac{1}{2} \mathbf{y}^T \mathbf{A} \mathbf{y} \\
&= -w_2(\mathbf{b}) + A_2(\mathbf{y})
\end{aligned}$$

where

$$w_2(\mathbf{b}) = \frac{1}{2} \mathbf{b}^T \mathbf{A}^{-1} \mathbf{b} \quad (1.7)$$

Hence,

$$\begin{aligned}
\mathcal{Z}(\mathbf{A}, \mathbf{b}) &= e^{w_2(\mathbf{b})} \int d^n y e^{-A_2(\mathbf{y})} \\
&= e^{w_2(\mathbf{b})} \mathcal{Z}(\mathbf{A}) \\
&= \frac{(2\pi)^{n/2}}{\sqrt{\det \mathbf{A}}} \exp\left(\frac{1}{2} \mathbf{b}^T \mathbf{A}^{-1} \mathbf{b}\right)
\end{aligned} \quad (1.8)$$

Note that eq(1.8) is applicable only if  $\mathbf{A}$  is real, symmetric & positive definite.

## Gaussian Expectation Values

$$\langle x_{k_1} x_{k_2} \dots x_{k_l} \rangle \equiv \frac{1}{\mathcal{Z}(\mathbf{A})} \int d^n x e^{-A_2(\mathbf{x})} x_{k_1} x_{k_2} \dots x_{k_l} \quad (1.9)$$

so that

$$\begin{aligned}
\langle 1 \rangle &\equiv \frac{1}{\mathcal{Z}(\mathbf{A})} \int d^n x e^{-A_2(\mathbf{x})} = 1 \\
\mathcal{Z}(\mathbf{A}, \mathbf{b}) &= \int d^n x e^{-A_2(\mathbf{x}) + \mathbf{b}^T \mathbf{x}} \\
\rightarrow \frac{\partial}{\partial b_k} \mathcal{Z}(\mathbf{A}, \mathbf{b}) &= \int d^n x e^{-A_2(\mathbf{x}) + \mathbf{b}^T \mathbf{x}} x_k \\
\therefore \langle x_k \rangle &= \frac{1}{\mathcal{Z}(\mathbf{A})} \frac{\partial}{\partial b_k} \mathcal{Z}(\mathbf{A}, \mathbf{b}) \Big|_{\mathbf{b}=0} \\
&= \frac{\sqrt{\det \mathbf{A}}}{(2\pi)^{n/2}} \frac{\partial}{\partial b_k} \mathcal{Z}(\mathbf{A}, \mathbf{b}) \Big|_{\mathbf{b}=0}
\end{aligned}$$

Similarly,

$$\begin{aligned}
\langle x_{k_1} x_{k_2} \dots x_{k_l} \rangle &= \frac{1}{\mathcal{Z}(\mathbf{A})} \left[ \frac{\partial}{\partial b_{k_1}} \dots \frac{\partial}{\partial b_{k_l}} \mathcal{Z}(\mathbf{A}, \mathbf{b}) \right] \Big|_{\mathbf{b}=0} \\
&= \frac{\sqrt{\det \mathbf{A}}}{(2\pi)^{n/2}} \left[ \frac{\partial}{\partial b_{k_1}} \dots \frac{\partial}{\partial b_{k_l}} \mathcal{Z}(\mathbf{A}, \mathbf{b}) \right] \Big|_{\mathbf{b}=0}
\end{aligned} \quad (1.10)$$

$$\mathcal{Z}(\mathbf{A}, \mathbf{b}) = e^{w_2(\mathbf{b})} \mathcal{Z}(\mathbf{A})$$

$$\rightarrow \langle x_{k_1} x_{k_2} \dots x_{k_l} \rangle = \left[ \frac{\partial}{\partial b_{k_1}} \dots \frac{\partial}{\partial b_{k_l}} e^{w_2(\mathbf{b})} \right] \Big|_{\mathbf{b}=0} \quad (1.11)$$

For any analytic function

$$F(\mathbf{x}) = \sum_{k=0}^{\infty} \left( \frac{d^k F}{d x^k} \right)_{x=0} \frac{x^k}{k!}$$

we have

$$\begin{aligned} \langle F(\mathbf{x}) \rangle &= \sum_{k=0}^{\infty} \left( \frac{d^k F}{d x^k} \right)_{x=0} \frac{\langle x^k \rangle}{k!} \\ &= \sum_{k=0}^{\infty} \left( \frac{d^k F}{d x^k} \right)_{x=0} \frac{1}{k!} \left( \frac{\partial}{\partial b} \right)^k e^{w_2(\mathbf{b})} \Big|_{\mathbf{b}=0} \\ &= F \left( \frac{\partial}{\partial b} \right) e^{w_2(\mathbf{b})} \Big|_{\mathbf{b}=0} \end{aligned} \quad (1.12)$$

### Wick's Theorem

$$\begin{aligned} w_2(\mathbf{b}) &= \frac{1}{2} \mathbf{b}^T \mathbf{A}^{-1} \mathbf{b} = \frac{1}{2} \sum_{i,j=1}^n b_i A_{ij}^{-1} b_j \\ \rightarrow \frac{\partial}{\partial b_k} w_2(\mathbf{b}) &= \frac{1}{2} \left( \sum_{j=1}^n A_{kj}^{-1} b_j + \sum_{i=1}^n b_i A_{ik}^{-1} \right) \\ &= \sum_{j=1}^n A_{kj}^{-1} b_j \quad (\mathbf{A} \text{ is symmetric}) \\ \therefore \frac{\partial}{\partial b_k} e^{w_2(\mathbf{b})} &= \sum_{k'=1}^n A_{kk'}^{-1} b_{k'} e^{w_2(\mathbf{b})} \\ \langle x_k \rangle &= \sum_{k'=1}^n A_{kk'}^{-1} b_{k'} e^{w_2(\mathbf{b})} \Big|_{\mathbf{b}=0} = 0 \\ \frac{\partial}{\partial b_j} \frac{\partial}{\partial b_k} e^{w_2(\mathbf{b})} &= \frac{\partial}{\partial b_j} \sum_{k'=1}^n A_{kk'}^{-1} b_{k'} e^{w_2(\mathbf{b})} \\ &= \left( A_{kj}^{-1} + \sum_{k'=1}^n A_{kk'}^{-1} b_{k'} \sum_{j'=1}^n A_{j'j}^{-1} b_{j'} \right) e^{w_2(\mathbf{b})} \\ \rightarrow \langle x_j x_k \rangle &= A_{kj}^{-1} = A_{jk}^{-1} \\ \frac{\partial}{\partial b_m} \frac{\partial}{\partial b_j} \frac{\partial}{\partial b_k} e^{w_2(\mathbf{b})} &= \left[ \left( A_{kj}^{-1} + \sum_{k'=1}^n A_{kk'}^{-1} b_{k'} \sum_{j'=1}^n A_{j'j}^{-1} b_{j'} \right) \sum_{m'=1}^n A_{m'm}^{-1} b_{m'} \right. \\ &\quad \left. + A_{km}^{-1} \sum_{j'=1}^n A_{j'j}^{-1} b_{j'} + \sum_{k'=1}^n A_{kk'}^{-1} b_{k'} A_{jm}^{-1} \right] e^{w_2(\mathbf{b})} \\ \rightarrow \langle x_m x_j x_k \rangle &= 0 \end{aligned}$$

$$\begin{aligned} \frac{\partial}{\partial b_p} \frac{\partial}{\partial b_m} \frac{\partial}{\partial b_j} \frac{\partial}{\partial b_k} e^{w_2(b)} = & \left\{ \left[ \left( A_{kj}^{-1} + \sum_{k'=1}^n A_{kk'}^{-1} b_{k'} \sum_{j'=1}^n A_{jj'}^{-1} b_{j'} \right) \sum_{m'=1}^n A_{mm'}^{-1} b_{m'} \right. \right. \\ & + A_{km}^{-1} \sum_{j'=1}^n A_{jj'}^{-1} b_{j'} + \sum_{k'=1}^n A_{kk'}^{-1} b_{k'} A_{jm}^{-1} \left. \right] \sum_{p'=1}^n A_{pp'}^{-1} b_{p'} \\ & + \left( A_{kp}^{-1} \sum_{j'=1}^n A_{jj'}^{-1} b_{j'} + \sum_{k'=1}^n A_{kk'}^{-1} b_{k'} A_{jp'}^{-1} \right) \sum_{m'=1}^n A_{mm'}^{-1} b_{m'} \\ & + \left( A_{kj}^{-1} + \sum_{k'=1}^n A_{kk'}^{-1} b_{k'} \sum_{j'=1}^n A_{jj'}^{-1} b_{j'} \right) A_{mp}^{-1} \\ & \left. + A_{km}^{-1} A_{jp}^{-1} + A_{kp}^{-1} A_{jm}^{-1} \right\} e^{w_2(b)} \end{aligned}$$

$$\begin{aligned} \rightarrow \langle x_p x_m x_j x_k \rangle &= A_{kj}^{-1} A_{mp}^{-1} + A_{km}^{-1} A_{jp}^{-1} + A_{kp}^{-1} A_{jm}^{-1} \\ &= \langle x_k x_j \rangle \langle x_m x_p \rangle + \langle x_k x_m \rangle \langle x_j x_p \rangle + \langle x_k x_p \rangle \langle x_j x_m \rangle \\ &= \sum_{\{a,b,c,d\}} A_{ab}^{-1} A_{cd}^{-1} \\ &= \sum_{\{a,b,c,d\}} \langle x_a x_b \rangle \langle x_c x_d \rangle \end{aligned}$$

where  $\{a, b, c, d\} =$  All possible pairings of  $\{p m j k\}$ .

Continuing the process, we see that

$$\begin{aligned} \langle x_{k_1} \dots x_{k_m} \rangle &= 0 & \text{if } m &= \text{odd} \\ \langle x_{k_1} \dots x_{k_m} \rangle &= \sum_{\{a_1, b_1, \dots, a_{m/2}, b_{m/2}\}} A_{a_1 b_1}^{-1} \dots A_{a_{m/2} b_{m/2}}^{-1} & \text{if } m &= \text{even} \\ &= \sum_{\{a_1, b_1, \dots, a_{m/2}, b_{m/2}\}} \langle x_{a_1} x_{b_1} \rangle \dots \langle x_{a_{m/2}} x_{b_{m/2}} \rangle \end{aligned}$$

where

$$\{a_1, b_1 \dots a_{m/2}, b_{m/2}\} = \text{All possible pairings of } \{k_1 \dots k_m\}.$$

which is just the Wick's theorem.