

~~Lecture 6~~: } IBP. Dyson-Schwinger etc.
Wed 1st lecture } prepare for small scale limit

First: Recall Wick (Isserlis) Theorem:

Given a mean-zero multivar Gaussian $\{\Phi(x)\}_{x \in \Lambda}$.

e.g: Φ is massive GFF on $\Lambda_{\varepsilon, m}$ with cov $(m - \Delta_\varepsilon)^{-1}$ as in previous lectures.

$$\text{Then } \mathbb{E}[\Phi(x_1) \dots \Phi(x_n)] = \begin{cases} 0 & \text{if } n \text{ odd} \\ \sum_p \prod_{\{i,j\} \in p} \underbrace{\mathbb{E}[\Phi(x_i) \Phi(x_j)]}_{=\text{cov}} & \text{if } n \text{ even} \end{cases}$$

Here $\sum_p$ is over all pairings p of $\{1, \dots, n\}$
i.e. partitions of $\{1, \dots, n\}$ into pairs $\{i, j\}$

Example: $\mathbb{E}[\Phi(x_1) \dots \Phi(x_4)] = \mathbb{E}[\Phi(x_1) \Phi(x_2)] \mathbb{E}[\Phi(x_3) \Phi(x_4)]$
 $+ \mathbb{E}[\Phi(x_1) \Phi(x_3)] \mathbb{E}[\Phi(x_2) \Phi(x_4)]$
 $+ \mathbb{E}[\Phi(x_1) \Phi(x_4)] \mathbb{E}[\Phi(x_2) \Phi(x_3)]$

Wick theorem says:

For Gaussian, n -point correlations $\rightarrow$ 2-point correlations
(and perhaps mean)

Not true for Non-Gaussian e.g. Φ^4 but there are
"Dyson-Schwinger" equations which relate
correlations of different orders.

Fix finite lattice.

$$\begin{aligned}
 \nu = \nu_{\varepsilon, m} &= e^{-\varepsilon^d \int_x \left\{ (\nabla \cdot \varphi)^2 + \frac{m^2}{2} \varphi^2 + \frac{\lambda}{4} \varphi^4 \right\}} \prod_x d\varphi_x / Z \\
 &= e^{-V(\varphi)} \prod_x d\varphi_x / Z
 \end{aligned}$$

For function $F(\varphi)$ write gradient: $DF(\varphi) = \left(\frac{\partial F}{\partial \varphi_x} \right)_{x \in \Lambda}$

Fix x

$$\begin{aligned}
 \int_{\mathbb{R}^N} DF(\varphi)(x) \nu(d\varphi) &= \int \frac{\partial F(\varphi)}{\partial \varphi_x} \nu(d\varphi) \quad \text{IBP for Leb measure} \\
 &= - \int F(\varphi) \cdot \left(\frac{\partial}{\partial \varphi_x} e^{-V} \right) \cdot \prod_x d\varphi_x / Z \\
 &= \int F(\varphi) \frac{\partial V}{\partial \varphi_x} \cdot \nu(d\varphi) \\
 &= \int F(\varphi) \left((m^2 - \Delta) \varphi_x + \lambda \varphi_x^3 \right) \nu(d\varphi)
 \end{aligned}$$

This is IBP for ν .

"under $\mathbb{E}_\nu$, derivative is effectively same as multiply $(\dots)$ ".

Dyson-Schwinger:

Take $F(\varphi) = \varphi(x_1) \dots \varphi(x_n)$

$$DF(\varphi)(x) = \frac{\partial F(\varphi)}{\partial \varphi_x} = \sum_{i=1}^n \delta(x - x_i) \varphi(x_1) \dots \varphi(x_{i-1}) \varphi(x_{i+1}) \dots \varphi(x_n)$$

IBP $\Rightarrow$

$$\sum_{i=1}^n \delta(x - x_i) \mathbb{E}(\varphi(x_1) \dots \varphi(x_{i-1}) \varphi(x_{i+1}) \dots \varphi(x_n))$$

$n-1$ point correlation

$$= \mathbb{E} \left((m^2 - \Delta_x) \varphi(x) \varphi(x_1) \cdots \varphi(x_n) \right) \quad n+1 \text{ point correlation}$$

$$+ \lambda \mathbb{E} \left(\varphi^3(x) \varphi(x_1) \cdots \varphi(x_n) \right) \quad n+2 \text{ point correlation}$$

e.g; $n=1$

$$\delta(x-y) = (m^2 - \Delta_x) \mathbb{E}(\varphi_x \varphi_y) + \lambda \mathbb{E}(\varphi_x^3 \varphi_y)$$

In particular if $\lambda=0$, we recover:
for GFF, $\mathbb{E}(\varphi_x \varphi_y)$ is Green's function of $m^2 - \Delta$.

$n=3$:

$$\delta(x-x_1) \mathbb{E}(\varphi(x_2) \varphi(x_3)) + \delta(x-x_2) \mathbb{E}(\varphi(x_1) \varphi(x_3))$$

$$+ \delta(x-x_3) \mathbb{E}(\varphi(x_1) \varphi(x_2))$$

$$= \mathbb{E} \left((m^2 - \Delta_x) \varphi(x) \varphi(x_1) \cdots \varphi(x_3) \right) + \lambda \mathbb{E}(\varphi(x)^3 \varphi(x_1) \cdots \varphi(x_3))$$

Convolve with $C = (m - \Delta)^{-1}$ at x ;

$$C(x-x_1) \mathbb{E}(\varphi(x_2) \varphi(x_3)) + C(x-x_2) \mathbb{E}(\varphi(x_1) \varphi(x_3))$$

$$+ C(x-x_3) \mathbb{E}(\varphi(x_1) \varphi(x_2))$$

$$= \mathbb{E}(\varphi(x) \varphi(x_1) \varphi(x_2) \varphi(x_3)) + \lambda \sum_z C(x-z) \mathbb{E}(\varphi(z)^3 \varphi(x_1) \cdots \varphi(x_3))$$

In particular if $\lambda=0$, we recover Wick theorem (4 pt) for GFF.
higher moments Wick follow similarly.

Note: for non-Gaussian, DS eqs don't close.

Next time, we'll discuss small scale limit $\varepsilon \rightarrow 0$.

$$Z_{\varepsilon, M} = e^{-\varepsilon^d \sum_x \left\{ \left(\nabla_\varepsilon \phi \right)^2 + \frac{m^2}{2} \phi^2 + \frac{\lambda}{6} \phi^3 \right\}} \prod_x d\phi_x / Z$$

Fix $M=1$

To prepare for it, Let's clarify scaling first:

$$\text{Recall } \nabla_\varepsilon \phi(x) = \frac{\phi(x+\varepsilon) - \phi(x)}{\varepsilon}$$

Dynamics:

$$\frac{d\phi}{dt} = \varepsilon^d (\Delta_\varepsilon \phi - m^2 \phi - \lambda \phi^3) + \frac{dB}{dt}$$

where $B = (B(x))_{x \in (\mathbb{Z})^d}$
indep
Standard BM's

$$\text{Let } t = \varepsilon^{-d} \tilde{t}$$

$$\frac{d\phi}{d\tilde{t}} = \Delta_\varepsilon \phi - m^2 \phi - \lambda \phi^3 + \frac{dB(\varepsilon^{-d} \tilde{t})}{d\tilde{t}}$$

$$\text{Recall: } \varepsilon^{d/2} B(\varepsilon^{-d} \tilde{t}) \stackrel{\text{law}}{=} B(\tilde{t})$$

$$\frac{d\phi}{d\tilde{t}} = \Delta_\varepsilon \phi - m^2 \phi - \lambda \phi^3 + \underbrace{\frac{dB}{d\tilde{t}} \cdot \varepsilon^{-d/2}}_{\downarrow}$$

Will converge in law to "space time white noise"

Def: $\mathbb{Z}$ centered Gaussian. $\mathbb{E}[\mathbb{Z}(f) \mathbb{Z}(g)] = \iint f g \, dt \, dx$

$$\left(\frac{dB}{d\tilde{t}} \cdot \varepsilon^{-d/2} \right)(f) = \varepsilon^d \sum_x \int \varepsilon^{-d/2} \frac{dB}{d\tilde{t}}(t, x) f(t, x) \, dt$$

$$\text{So, } \mathbb{E} \left[\varepsilon^{2d} \sum_{x, y} \cdot \varepsilon^{-d} \iint \frac{dB}{d\tilde{t}}(t, x) \frac{dB}{d\tilde{t}}(s, y) f(t, x) g(s, y) \, dt \, ds \right]$$

$$\stackrel{\text{Itô}}{=} \varepsilon^d \sum_x \int f(t, x) g(t, x) \, dt \rightarrow \iint f g \, dt \, dx$$

Below it will be more convenient to consider

$$\frac{d\phi}{dt} = \Delta_\varepsilon \phi - m^2 \phi - \lambda \phi^4 + \tilde{Z}_\varepsilon \quad \text{where } \tilde{Z}_\varepsilon \rightarrow \tilde{Z} \text{ a.s.}$$

for example $\tilde{Z}_\varepsilon$ is directly defined from $\tilde{Z}$ by "discretization":

$$\text{e.g. } \tilde{Z}_\varepsilon(t, x) = \varepsilon^{-d} \langle \tilde{Z}(t, \cdot), \mathbb{1}_{| \cdot - x | \leq \frac{\varepsilon}{2}} \rangle_{L^2(\mathbb{R}^d)}$$

$$\text{Exercise: } \tilde{Z}_\varepsilon \stackrel{\text{law}}{=} \varepsilon^{-\frac{d}{2}} \frac{dB}{dt}$$

Two alternative views of $\tilde{Z}$ (useful in calculations)

$$1) \mathbb{E}[\tilde{Z}(t, x) \tilde{Z}(t', x')] = \delta(t - t') \delta(x - x')$$

$$\left(\begin{aligned} \mathbb{E}[\tilde{Z}(f) \tilde{Z}(g)] &= \mathbb{E} \left[\int \tilde{Z}(t, x) \tilde{Z}(t', x') f(t, x) g(t', x') \frac{dt dx dt' dx'}{dx dx'} \right] \\ &= \int \delta(t - t') \delta(x - x') f(t, x) g(t', x') dt dt' dx dx' \\ &= \int f(t, x) g(t, x) dt dx \end{aligned} \right)$$

$$2) \tilde{Z}(t) = \sum_k \hat{\tilde{Z}}(t, k) e^{ix \cdot k}$$

$$\text{where } \hat{\tilde{Z}}(t, k) = \frac{1}{i2} \left(\frac{dB_k}{dt} + i \frac{dB'_k}{dt} \right)$$

$$\hat{\tilde{Z}}(t, -k) = \overline{\hat{\tilde{Z}}(t, k)}$$

$(B_k, B'_k)_{k \in \mathbb{Z}^d \cap [-\frac{1}{\varepsilon}, \frac{1}{\varepsilon}]}$ are indep standard BM

$$\text{Exercise: } \mathbb{E}[\hat{\tilde{Z}}(t, k) \overline{\hat{\tilde{Z}}(t', k')}] = 2\pi \mathbb{1}_{k=k'} \delta(t - t')$$

$$\left(\begin{aligned} \text{So, } \mathbb{E}[\tilde{Z}(t, x) \tilde{Z}(t', x')] &= \mathbb{E} \sum_{k, k'} \hat{\tilde{Z}}(t, k) \hat{\tilde{Z}}(t', k') e^{ix \cdot k} e^{ix' \cdot k'} \\ &= \delta(t - t') \delta(x - x') \quad \text{since } \sum_k e^{i(x-x') \cdot k} = \delta(x - x') \end{aligned} \right)$$

For Friday lectures,
the following fact is important:

★ $Y(t)$ is function if $d=1$. $Y(t)$ is distribution if $d \geq 2$.

We prove exact regularity next time, but for now we put
in simpler way:

$$\mathbb{E}(Y(t, x)^2) = \begin{cases} < \infty & d=1 \\ = \infty & d \geq 2 \end{cases}$$

Indeed, $Y(t, x) = \int_{-\infty}^t \int_{\mathbb{R}^d} P_m(t-s, x-y) \xi(s, y) ds dy$

$$\begin{aligned} \mathbb{E}(Y(t, x)^2) &\stackrel{\text{It\^o}}{=} \int_{-\infty}^t \int_{\mathbb{R}^d} P_m(t-s, x-y)^2 ds dy \\ &\stackrel{\text{Parseval}}{=} \int_{-\infty}^t \int_{\mathbb{R}^d} e^{-(t-s)(|k|^2+m^2)} dk ds = \int_{\mathbb{R}^d} \frac{1}{|k|^2+m^2} dk = \begin{cases} < \infty & d=1 \\ = \infty & d \geq 2 \end{cases} \end{aligned}$$

understood as γ_s and $s \rightarrow 0$

But Y is distribution since $\forall \varphi \in \mathcal{S}(\mathbb{R}^d)$,

$$\mathbb{E}\left[\left(\int Y(t, x) \varphi(x) dx\right)^2\right] = \iint \varphi(x) \varphi(y) \mathbb{E}(Y(t, x) Y(t, y)) dx dy$$

$$= \iint \varphi(x) \varphi(y) \int_{-\infty}^t \int_{\mathbb{R}^d} P_m(t-s, x-z) P_m(t-s, y-z) dz ds < \infty$$

one way to see is $= \sum_k |\hat{\varphi}(k)|^2 \frac{1}{|k|^2+m^2} < \infty$

Another way:

$$\left| \int_{-\infty}^t \int_{\mathbb{R}^d} P_m(t-s, x-z) P_m(t-s, y-z) dz ds \right| \lesssim \frac{1}{|x-y|^{d-2-\varepsilon}} \text{ for } |x-y| \text{ small.}$$

exercise: " $-d - d + (d+2) = -d+2$ " (maybe TA) always integrable around 0.