

Séminaire du LJLL, Paris

Metastability of Allen-Cahn equations with weak space-time white noise

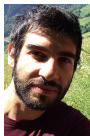
Nils Berglund

Institut Denis Poisson, University of Orléans, France



9 January 2026

Joint works with Barbara Gentz (Bielefeld), Giacomo Di Gesù (Rome),
Hendrik Weber (Münster), Tom Klose (Oxford) and Nikolas Tapia (Berlin)



Plan

1. Metastability in Allen–Cahn equations on \mathbb{T}^1

Joint work with Barbara Gentz

Electronic J. Probability **18**, (24):1–58 (2013)

2. Metastability in Allen–Cahn equations on \mathbb{T}^2

Joint work with Giacomo Di Gesù & Hendrik Weber

Electronic J. Probability **22**, (41):1–27 (2017)

3. Current work, motivated by the case of \mathbb{T}^3

Joint works with Tom Klose and Nikolas Tapia

[arXiv/2207.08555](https://arxiv.org/abs/2207.08555) (2022) (Proceedings, Vienna trimester)

[arXiv/2507.03820](https://arxiv.org/abs/2507.03820) (2025)

Allen–Cahn equation on \mathbb{T}^2

$$\partial_t \phi(t, x) = \nu \Delta \phi(t, x) + \phi(t, x) - \phi(t, x)^3 \quad (\nu = \nu(\varepsilon t))$$

(Online: <https://youtu.be/yXOEAxZHNCQ>)

1. Deterministic Allen–Cahn PDE in $d = 1$

[Chafee & Infante 74, Allen & Cahn 75]

$$\partial_t \phi(t, x) = \Delta \phi(t, x) + f(\phi(t, x))$$

- ▷ $x \in [0, L]$, L : bifurcation parameter ($x \mapsto L^{-1}x \Rightarrow \Delta \mapsto \nu \Delta$, $\nu = L^{-2}$)
- ▷ $\phi(t, x) \in \mathbb{R}$
- ▷ Either periodic or zero-flux Neumann boundary conditions
- ▷ In this talk: $f(\phi) = \phi - \phi^3$ (results more general)

Energy function:

$$V[\phi] = \int_0^L \left[\frac{1}{2} \phi'(x)^2 - \frac{1}{2} \phi(x)^2 + \frac{1}{4} \phi(x)^4 \right] dx \quad \Rightarrow \quad \partial_t \langle \phi, \psi \rangle = -\nabla_\psi V(\phi)$$

Stationary solutions: $\phi_0''(x) = -\phi_0(x) + \phi_0(x)^3$ critical points of V

Stability: Sturm–Liouville problem $\partial_t \psi_t(x) = \psi_t''(x) + [1 - 3\phi_0(x)^2] \psi_t(x)$

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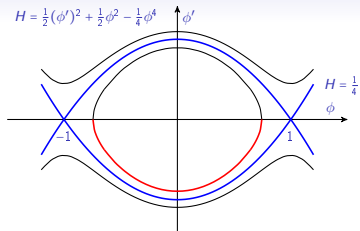
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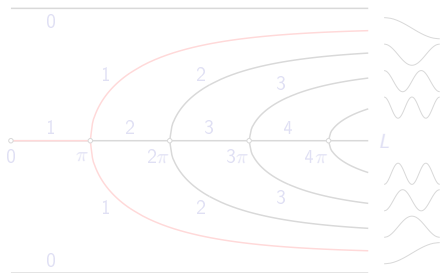
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- ▷ $\phi_{\pm}(x) \equiv \pm 1$: **stable**
- ▷ $\phi_0(x) \equiv 0$: **unstable**
- ▷ Nonconstant solutions satisfying b.c.
(expressible in terms of Jacobi elliptic fcts)
- ▷ Neumann b.c: $2k$ nonconstant solutions when $L > k\pi$



Number of positive
eigenvalues
(= unstable directions)
Transition state

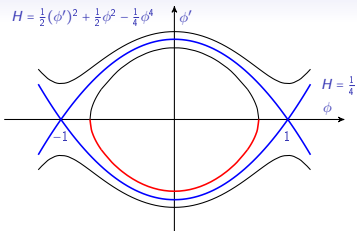


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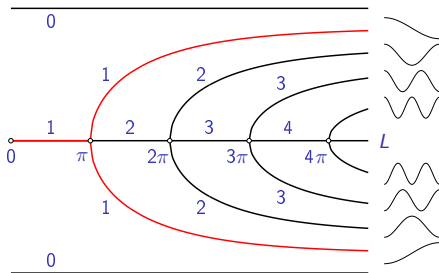
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Allen–Cahn SPDE with space-time white noise

$$\partial_t \phi(t, x) = \Delta \phi(t, x) + f(\phi(t, x)) + \sqrt{2\varepsilon} \xi(t, x) \quad (f(\phi) = \phi - \phi^3)$$

- ▷ Space-time white noise: ξ Gaussian random distribution,
Formal def: $\mathbb{E}[\xi(t, x)] = 0$, $\mathbb{E}[\xi(t, x)\xi(s, y)] = \delta(t - s)\delta(x - y)$
Rigorous def: $\mathbb{E}[(\xi, \varphi_1)(\xi, \varphi_2)] = \langle \varphi_1, \varphi_2 \rangle_{L^2}$ \forall test functions φ_1, φ_2
In Fourier basis: $\xi(t, x) = \sum_{k \in \mathbb{Z}} \dot{W}_k(t) e_k(x)$, $W_k(t)$ independent BM
- ▷ Local solution theory: use Duhamel principle
for PDE $\partial_t \phi = \mathcal{L}\phi + f(\phi) + \sqrt{2\varepsilon} \xi$ where
 $\mathcal{L} = \Delta + f'(\phi)$ is linear
▷ \mathcal{L} is elliptic \Rightarrow \mathcal{L}^{-1} is bounded
▷ \mathcal{L} is dissipative \Rightarrow \mathcal{L}^{-1} is contractive

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Facts: $(P * \xi)(t, \cdot) \in H^s(\mathbb{T}) \quad \forall s < \frac{1}{2}$ and $P * \xi \in \mathcal{C}_s^\alpha(\mathbb{R}_+ \times \mathbb{T}) \quad \forall \alpha < \frac{1}{2}$
- ▶ **Global solution theory:** use energy estimates

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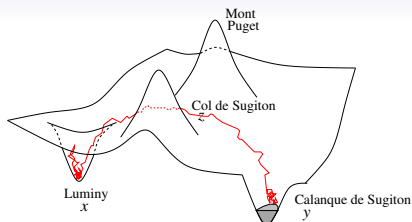
Reversible diffusion in a double-well

$$dx_t = -\nabla V(x_t) dt + \sqrt{2\varepsilon} dW_t$$

$V : \mathbb{R}^d \rightarrow \mathbb{R}$ confining potential

$$\tau_y^x = \inf\{t > 0 : x_t \in \mathcal{B}_\varepsilon(y)\}$$

first-hitting time of small ball $\mathcal{B}_\varepsilon(y)$,
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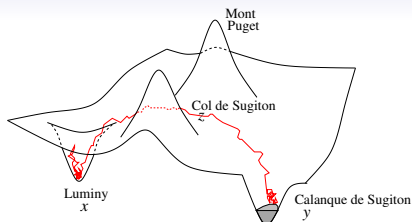
Arrhenius' law (1889): $\mathbb{E}[\tau_y^x] \simeq e^{[V(z)-V(x)]/\varepsilon}$

Eyring–Kramers law (1935, 1940):

Eigenvalues of Hessian of V at minimum x : $0 < \nu_1 \leq \nu_2 \leq \dots \leq \nu_d$

Eigenvalues of Hessian of V at saddle z : $\lambda_1 < 0 < \lambda_2 \leq \dots \leq \lambda_d$

$$\mathbb{E}[\tau_y^x] = 2\pi \sqrt{\frac{\lambda_2 \dots \lambda_d}{|\lambda_1| \nu_1 \dots \nu_d}} e^{[V(z)-V(x)]/\varepsilon} [1 + \mathcal{O}_\varepsilon(1)]$$



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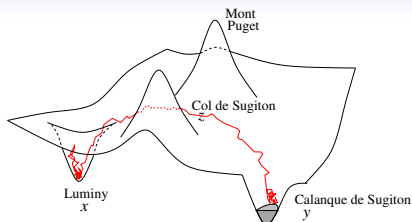
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Arrhenius' law: proved by [Freidlin, Wentzell, 1979] using large deviations

Eyring–Kramers law: [Bovier, Eckhoff, Gayard, Klein, 2004] using potential theory, [Helffer, Klein, Nier, 2004] using Witten Laplacian, ...



Potential-theoretic proof of Eyring–Kramers law

$$\triangleright w_A(x) = \mathbb{E}^x[\tau_A] \quad \text{satisfies} \quad \begin{cases} (\mathcal{L} w_A)(x) = -1 & x \in A^c \\ w_A(x) = 0 & x \in A \end{cases}$$

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Theorem: $A, B \subset \mathbb{R}^d$ disjoint. \exists proba measure ν_{AB} on ∂A s.t.

$$\int_{\partial A} \mathbb{E}^x[\tau_B] \nu_{AB}(dx) = \frac{1}{\text{cap}(A, B)} \int_{B^c} e^{-V(y)/\varepsilon} h_{AB}(y) dy$$

Apply to A, B neighbourhoods of x, y

$$\triangleright \text{Laplace asymptotics: } \int_{B^c} h_{A,B}(y) e^{-V(y)/\varepsilon} dy \simeq \sqrt{\frac{(2\pi\varepsilon)^d}{\nu_1 \dots \nu_d}} e^{-V(x)/\varepsilon}$$

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Potential-theoretic proof of Eyring–Kramers law

Theorem: Dirichlet principle

Let $\mathcal{H}_{AB} = \{h: \mathbb{R}^d \rightarrow [0, 1]: h|_A = 1, h|_B = 0\}$. Then

$$\text{cap}(A, B) = \inf_{h \in \mathcal{H}_{AB}} \mathcal{E}(h) = \mathcal{E}(h_{AB})$$

Appropriate test function yields $\text{cap}(A, B) \lesssim \varepsilon \sqrt{\frac{|\lambda_1|}{2\pi\varepsilon}} \sqrt{\frac{(2\pi\varepsilon)^{d-1}}{\lambda_2 \dots \lambda_d}} e^{-V(z)/\varepsilon}$

Theorem: Thomson principle [Landim, Mariani, Seo 2018]

Let $\mathcal{U}_{AB} = \{f: \nabla \cdot f|_{(A \cup B)^c} = 0, \int_{\partial A} f(x) \cdot n_A(x) \sigma(dx) = 1\}$. Then

$$\text{cap}(A, B) = \sup_{f \in \mathcal{U}_{AB}} \frac{1}{\mathcal{D}(f)} = \frac{1}{\mathcal{D}(f_{AB})} \quad \mathcal{D}(f) = \frac{1}{\varepsilon} \int e^{V(x)/\varepsilon} |f(x)|^2 dx$$

Appropriate test flow yields $\text{cap}(A, B) \gtrsim \varepsilon \sqrt{\frac{|\lambda_1|}{2\pi\varepsilon}} \sqrt{\frac{(2\pi\varepsilon)^{d-1}}{\lambda_2 \dots \lambda_d}} e^{-V(z)/\varepsilon}$

$$\Rightarrow \mathbb{E}^x[\tau_B] \simeq \int_{\partial A} \mathbb{E}^x[\tau_B] \nu_{AB}(dx) \simeq 2\pi \sqrt{\frac{\lambda_2 \dots \lambda_d}{|\lambda_1| \nu_1 \dots \nu_d}} e^{-[V(z) - V(x)]/\varepsilon}$$

Potential-theoretic proof of Eyring–Kramers law

Theorem: Dirichlet principle

Let $\mathcal{H}_{AB} = \{h : \mathbb{R}^d \rightarrow [0, 1] : h|_A = 1, h|_B = 0\}$. Then

$$\text{cap}(A, B) = \inf_{h \in \mathcal{H}_{AB}} \mathcal{E}(h) = \mathcal{E}(h_{AB})$$

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Theorem: Thomson principle [Landim, Mariani, Seo 2018]

Let $\mathcal{U}_{AB} = \{f : \nabla \cdot f|_{(A \cup B)^c} = 0, \int_{\partial A} f(x) \cdot n_A(x) \sigma(dx) = 1\}$. Then

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Eyring–Kramers law for 1D SPDEs: heuristics

$$\partial_t \phi(t, x) = \Delta \phi(t, x) + f(\phi(t, x)) + \sqrt{2\varepsilon} \xi(t, x) \quad (f(\phi) = \phi - \phi^3)$$

Initial condition: ϕ_{in} near $\phi_- \equiv -1$ with eigenvalues $\nu_k = \left(\frac{\beta k \pi}{L}\right)^2 + 2$

Target: $\phi_+ \equiv 1$, $\tau_+ = \inf\{t > 0: \|\phi(t, \cdot) - \phi_+\|_{L^\infty} < \rho\}$

Transition state: ($\beta = 1$ for Neumann b.c., $\beta = 2$ for periodic b.c.)

$$\phi_{\text{ts}}(x) = \begin{cases} \phi_0(x) \equiv 0 & \text{if } L \leq \beta\pi \quad \text{with ev } \lambda_k = \left(\frac{\beta k \pi}{L}\right)^2 - 1 \\ \phi_1(x) \text{ } \beta\text{-kink stationary sol.} & \text{if } L > \beta\pi \quad \text{with ev } \lambda'_k \end{cases}$$

[Faris & Jona-Lasinio 82]: large-deviation principle

\Rightarrow Arrhenius law: $\mathbb{E}^{\phi_{\text{in}}}[\tau_+] \simeq e^{(V[\phi_{\text{ts}}] - V[\phi_-])/\varepsilon}$

[Maier & Stein 01]: formal computation; for Neumann b.c. and $L < \pi$

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Eyring–Kramers law for 1D SPDEs: main result

Theorem: Neumann b.c. [B & Gentz, 2013]

▷ If $L < \pi - c$ with $c > 0$, then

$$\mathbb{E}^{\phi_{\text{in}}}[\tau_+] = 2\pi \sqrt{\frac{1}{|\lambda_0| \nu_0} \prod_{k=1}^{\infty} \frac{\lambda_k}{\nu_k}} e^{(V[\phi_{\text{ts}}] - V[\phi_-])/\varepsilon} [1 + \mathcal{O}(\varepsilon)]$$

▷ If $L > \pi + c$, then same formula with extra factor $\frac{1}{2}$ (since 2 saddles) and λ'_k instead of λ_k

▷ Results also for L near π and periodic b.c.

▷ Proof uses spectral Galerkin approximation & passage to the limit

▷ Prefactor involves a Fredholm determinant:

Δ_{\perp} Laplacian acting on mean zero functions

$$\prod_{k=1}^{\infty} \frac{\lambda_k}{\nu_k} = \det[(-\Delta_{\perp} - 1)(-\Delta_{\perp} + 2)^{-1}] = \det[\mathbb{1} - 3(-\Delta_{\perp} + 2)^{-1}]$$

converges because $\log \det = \text{Tr} \log$ and $(-\Delta_{\perp} + 2)^{-1}$ is trace class

$$\left(\text{limit} = \frac{\sqrt{2} \sin(L)}{\sinh(\sqrt{2}L)}\right)$$

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2. The two-dimensional case

- ▷ Large-deviation principle: [Hairer & Weber, 2015]
- ▷ Naive computation of prefactor fails:

$$\begin{aligned} \log \prod_{k \in (\mathbb{N}^2)^*} \frac{1 - \left(\frac{L}{|k|\pi}\right)^2}{1 + 2\left(\frac{L}{|k|\pi}\right)^2} &\simeq \sum_{k \in (\mathbb{N}^2)^*} \log \left(1 - \frac{3L^2}{|k|^2\pi^2}\right) \\ &\simeq - \sum_{k \in (\mathbb{N}^2)^*} \frac{3L^2}{|k|^2\pi^2} \simeq -\frac{3L^2}{\pi^2} \int_1^\infty \frac{r \, dr}{r^2} = -\infty \end{aligned}$$

- ▷ In fact, the equation needs to be renormalised ($P * \xi$ is distribution!)

Theorem: [Da Prato & Debussche 2003]

Let ξ^δ be a mollification on scale δ of white noise. Then

$$\partial_t \phi = \Delta \phi + \phi - [\phi^3 - 3\epsilon C(\delta)\phi] + \sqrt{2\epsilon} \xi^\delta$$

with $C(\delta) \simeq \log(\delta^{-1})$ admits local solution converging as $\delta \rightarrow 0$

Moreover, $\phi - \sqrt{2\epsilon}(P * \xi^\delta) \in \mathcal{C}([0, T], \mathcal{B}_{p,r}^\alpha) \cap L^p([0, T], \mathcal{B}_{p,r}^s)$

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Computation of the prefactor

- ▷ Consider for simplicity $L < \beta\pi \Rightarrow$ transition state in 0
- ▷ Galerkin-truncated renormalised potential

$$V_N = \frac{1}{2} \int_{\mathbb{T}^2} [\|\nabla\phi_N(x)\|^2 - \phi_N(x)^2] dx + \frac{1}{4} \int_{\mathbb{T}^2} :\phi_N(x)^4: dx$$

where $:\phi_N^4: = \phi_N^4 - 6\epsilon C_N \phi_N^2 + 3\epsilon^2 C_N^2$, $C_N = \sum_{|k| \leq N} \frac{1}{\lambda_k + 1} \asymp \log(N)$

- ▷ Using variational principles: $\text{cap}(A, B) \simeq \sqrt{\frac{|\lambda_0| \epsilon}{2\pi}} \prod_{0 < |k| \leq N} \sqrt{\frac{2\pi \epsilon}{\lambda_k}}$
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$$\int_{B^c} h_{A,B}(z) e^{-V_N(z)/\epsilon} dz = \frac{1}{2} \int e^{-V_N(z)/\epsilon} dz = \frac{1}{2} \mathcal{Z}_N(\epsilon)$$

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Main result in dimension 2

Theorem: [B, Di Gesù, Weber, 2017]

For $L < 2\pi$, $A \ni \phi_-$, $B \ni \phi_+$ balls in $\|\cdot\|_{H^s}$, $s < 0$, $\exists \mu_N$ prob measures on ∂A :

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with $T = 3(-\Delta_{\perp} - 1)^{-1}$

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3. Allen–Cahn equation in dimension 3

- ▷ Equation needs two counterterms:

$$\partial_t \phi = \Delta \phi + \phi - [\phi^3 - (3\varepsilon C_1(\delta) - 9\varepsilon^2 C_2(\delta))\phi] + \sqrt{2\varepsilon} \xi^\delta$$

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- ▷ Solution theories:
 - ◊ Regularity structures [Hairer 2014]
 - ◊ Paracontrolled distributions [Gubinelli–Imkeller–Perkowski 2015, Catellier & Chouk 2018]
 - ◊ Wilsonian RG [Kupiainen 2016]
- ▷ **Metastability**: one expects equivalent result (T is still Hilbert–Schmidt)
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with $C_1(\delta) \sim \delta^{-1}$, $C_2(\delta) \sim \log(\delta^{-1})$

- ▷ Solution theories:
 - ◊ Regularity structures [Hairer 2014]
 - ◊ Paracontrolled distributions [Gubinelli–Imkeller–Perkowski 2015, Catellier & Chouk 2018]
 - ◊ Wilsonian RG [Kupiainen 2016]
- ▷ **Metastability**: one expects equivalent result (T is still Hilbert–Schmidt)
- ▷ **Difficulty**: lower bound on capacity – finding a good test flow for Thomson principle
⇒ better understanding of expectations of polynomials under invariant measure

Φ_3^4 measure

- ▷ Potential (before renormalisation):

$$V_\alpha(\phi) = \int_\Lambda \left(\frac{1}{2} \|\nabla \phi(x)\|^2 + \frac{1}{2} \phi(x)^2 + \alpha \phi(x)^4 \right) dx$$

where $\Lambda = (\mathbb{R}/\mathbb{Z})^3 =: \mathbb{T}^3$, $\alpha \geq 0$

- ▷ Definition of Gibbs measure?

$$\nu_\alpha(d\phi) \text{ “=” } \frac{1}{Z_\alpha} e^{-V_\alpha(\phi)} \text{ “}d\phi\text{”}$$

- ▷ ν_0 is Gaussian (it's a GFF). For random variable F

$$\mathbb{E}^{\nu_\alpha}[F] = \frac{Z_0}{Z_\alpha} \mathbb{E}^{\nu_0}[F e^{-\alpha \int_\Lambda \phi^4(x) dx}]$$

In particular, $\frac{Z_\alpha}{Z_0} = \mathbb{E}^{\nu_0}[e^{-\alpha \int_\Lambda \phi^4(x) dx}]$

- ▷ Link with Allen–Cahn equation: write $\phi(x) = \phi_0 + \sqrt{\varepsilon} \phi_\perp(x)$, where ϕ_0 is average of ϕ (and $\alpha = \frac{\varepsilon}{4}$)

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The static Φ_3^4 model

Theorem: Potential needs exactly 4 counterterms:

$$V_\alpha(\phi) = \int_\Lambda \left(\frac{1}{2} \|\nabla \phi(x)\|^2 + \frac{1}{2} [1 - \alpha^2 C_N^{(2)}] \phi(x)^2 + \alpha : \phi(x)^4 :_{C_N^{(1)}} + \alpha^2 C_N^{(3)} - \alpha^3 C_N^{(4)} \right) dx$$

where N ($\sim \delta^{-1}$) UV cut-off and

$$C_N^{(1)} = G_N(0) = \text{Tr}((-\Delta_N + 1)^{-1}) = \mathcal{O}(N)$$

$$C_N^{(2)} = 4^2 3! \int_\Lambda G_N(x)^3 dx = \mathcal{O}(\log N)$$

$$C_N^{(3)} = 4^2 \frac{4!}{2!4^2} \int_\Lambda G_N(x)^4 dx = \mathcal{O}(N)$$

$$C_N^{(4)} = 4^3 \frac{2^3}{3!4^3} \binom{4}{2}^3 \int_\Lambda \int_\Lambda G_N(x)^2 G_N(y)^2 G_N(x-y)^2 dx dy = \mathcal{O}(\log N)$$

and $G_N(x) = \sum_{|k| \leq N} \frac{1}{\lambda_k + 1} e_k(x)$ is the Green function of Δ_N

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Perturbative computation of partition function

- ▷ Wick powers: $X = \text{X} = \int_{\Lambda} : \phi(x)^4 : dx$, $Y = \text{Y} = \int_{\Lambda} : \phi(x)^2 : dx$
- ▷ Potential: $V_{\alpha} = V_0 + \alpha X + \beta Y + \gamma$, $\beta = \frac{1}{2} \alpha^2 C_N^{(2)}$, $\gamma = \alpha^2 C_N^{(3)} - \alpha^3 C_N^{(4)}$
- Then $\frac{Z_{N,\alpha}}{Z_{N,0}} = \mathbb{E}^{\nu_0} [e^{-\alpha X - \beta Y - \gamma}] = e^{-\gamma} \mathbb{E}^{\nu_0} [e^{-\alpha X - \beta Y}]$
- ▷ Let $\Gamma = (\mathcal{V}, \mathcal{E})$ be a multigraph. Its valuation is

$$\Pi_N(\Gamma) = \int_{\Lambda^{\mathcal{V}}} \prod_{e \in \mathcal{E}} G_N(x_{e_+} - x_{e_-}) dx$$

For instance

$$C_N^{(1)} = \Pi_N \text{ (tadpole) } \text{ (diagram: a vertex with a loop attached to it)}$$

“tadpole”

$$C_N^{(2)} = 4^2 3! \Pi_N \text{ (sunset) } \text{ (diagram: two vertices connected by two edges)}$$

“sunset”

$$C_N^{(3)} = 4^2 \frac{4!}{2!4^2} \Pi_N \text{ (water melon) } \text{ (diagram: two vertices connected by three edges)}$$

“water melon”

$$C_N^{(4)} = 4^3 \frac{2^3}{3!4^3} \binom{4}{2} \Pi_N \text{ (PSG) } \text{ (diagram: three vertices forming a triangle with an edge from each vertex to the center)}$$

“PSG”

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Cumulant expansion

$$\triangleright \mathbb{E}^{\nu_0} [e^{-\alpha X - \beta Y}] = \sum_{n \geq 0} \frac{1}{n!} \mu_n$$

$$\mu_n = (-1)^n \mathbb{E}^{\nu_0} \left[\left(\alpha \begin{array}{c} \diagup \quad \diagdown \\ \times \end{array} + \beta \begin{array}{c} \bullet \\ \text{---} \end{array} \right)^n \right] = (-1)^n \sum_{m=0}^n \binom{n}{m} \alpha^m \beta^{n-m} A_{nm}$$

where $A_{nm} = \mathbb{E}^{\nu_0} \left[\begin{array}{c} \diagup \quad \diagdown \\ \times \end{array}^m \begin{array}{c} \bullet \\ \text{---} \end{array}^{n-m} \right]$ computed using Isserlis' thm

Examples: $\mu_2 = \alpha^2 4! \Pi_N \begin{array}{c} \curvearrowright \\ \curvearrowleft \end{array} + \beta^2 2! \Pi_N \begin{array}{c} \bullet \\ \text{---} \end{array}$

$$\mu_3 = -\alpha^3 \binom{4}{2}^3 2^3 \Pi_N \begin{array}{c} \triangle \\ \text{---} \end{array} - 3\alpha^2 \beta (4^2 \cdot 2 \cdot 3!) \Pi_N \begin{array}{c} \curvearrowright \\ \text{---} \\ \curvearrowleft \end{array} \\ - 3\alpha \beta^2 4! \Pi_N \begin{array}{c} \bullet \\ \text{---} \end{array} \begin{array}{c} \bullet \\ \text{---} \end{array} - 8\beta^3 \Pi_N \begin{array}{c} \triangle \\ \text{---} \end{array}$$

\triangleright Cumulant expansion: (Leonov & Shiraev)

$$-\log \mathbb{E}[e^{-\alpha X - \beta Y - \gamma}] = \gamma - \sum_{n=2}^{\infty} \frac{\kappa_n}{n!} \quad \kappa_n = \mu_n - \sum_{m=2}^{n-2} \binom{n-1}{m} \kappa_m \mu_{n-m}$$

\triangleright Linked Cluster Theorem: κ_n projection of μ_n on **connected** graphs

Proof: for instance Peccati & Taqqu (2011)

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Divergences and subdivergences

▷ Degree of Γ : $\deg(\Gamma) = 3(|\mathcal{V}| - 1) - |\mathcal{E}|$. Γ divergent if $\deg(\Gamma) \leq 0$.

▷ Examples:

$$\deg(\text{loop}) = 0$$

$$\Pi_N(\text{loop}) = \mathcal{O}(\log N)$$

$$\deg(\text{figure-eight}) = -1$$

$$\Pi_N(\text{figure-eight}) = \mathcal{O}(N)$$

$$\deg(\text{triangle}) = 0$$

$$\Pi_N(\text{triangle}) = \mathcal{O}(\log N)$$

▷ It looks like $\Pi_N(\Gamma) = \begin{cases} \mathcal{O}(N^{-\deg(\Gamma)}) & \text{if } \deg(\Gamma) < 0 \\ \mathcal{O}(\log N) & \text{if } \deg(\Gamma) = 0 \end{cases}$

However, $\deg(\text{figure-eight with top vertex}) = 1$, while $\Pi_N(\text{figure-eight with top vertex}) = \mathcal{O}(\log N)$
because it contains a subdivergence 

Theorem: [Bogolyubov, Parasiuk, Hepp, Zimmermann]

$\Pi_N^{\text{BPHZ}}(\Gamma) := -\Pi_N \mathcal{A}(\Gamma) \mathbf{1}_{\deg \Gamma > 0}$ is bdd uniformly in N for any Γ

where \mathcal{A} defined recursively by $\mathcal{A}(\Gamma) = -\Gamma - \sum_{\mathbf{1} \neq \bar{\Gamma} \not\subseteq \Gamma, \deg(\bar{\Gamma}) \leq 0} \mathcal{A}(\bar{\Gamma}) \cdot (\Gamma/\bar{\Gamma})$

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Divergences and subdivergences

- ▷ Degree of Γ : $\deg(\Gamma) = 3(|\mathcal{V}| - 1) - |\mathcal{E}|$. Γ divergent if $\deg(\Gamma) \leq 0$.
- ▷ Examples:

$$\deg(\text{bubble}) = 0$$

$$\Pi_N(\text{bubble}) = \mathcal{O}(\log N)$$


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Divergences and subdivergences

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$$\deg(\text{loop}) = 0$$

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However, $\deg(\text{triangle with subdivergence}) = 1$, while $\Pi_N(\text{triangle with subdivergence}) = \mathcal{O}(\log N)$

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Main result: Commutative diagram

$$\begin{array}{ccc}
 e^{-\alpha X} & \xrightarrow{\mathcal{P}} & \sum_p \frac{(-\alpha)^p}{p!} \mathcal{P}(X^p) \\
 \downarrow W & & \downarrow M^{\text{BPHZ}+\Theta} \\
 e^{-\alpha X - \beta Y} & \xrightarrow{\mathcal{P}} & \sum_{n,m} \frac{(-\alpha)^m (-\beta)^{n-m}}{m!(n-m)!} \mathcal{P}(X^m Y^{n-m}) \\
 & & \xrightarrow{\Pi_N} \log \mathbb{E}[e^{-\alpha X - \beta Y}]
 \end{array}$$

$\nwarrow \Pi_N^{\text{BPHZ}+\Pi_N\Theta}$

- ▷ $\mathcal{P} = \Pi_{\text{connected}}(\sum_{\text{pairings}})$, satisfies $\deg \mathcal{P}(X^n) = n - 3$
- ▷ W : Wick map, $W(X^n) = H_n(X; \beta Y)$ is Hermite polynomial (for more general models: Bell polynomial)
- ▷ Θ associated with energy renormalisation, $\Pi_N \Theta(e^{-\alpha X}) = \gamma$

Theorem: [B, Klose, Tapia 25]

$$\log \frac{Z_{N,\alpha}}{Z_{N,0}} = \log \mathbb{E}[e^{-\alpha X - \beta Y}] - \gamma \asymp - \sum_{n \geq 4} \frac{(-\alpha)^n}{n!} \Pi_N \mathcal{A}(\mathcal{P}(X^n))$$

as asymptotic expansion with terms uniformly bounded in N

Main result: Commutative diagram

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 & & \nwarrow \Pi_N^{\text{BPHZ}+\Pi_N\Theta}
 \end{array}$$

- ▷ $\mathcal{P} = \Pi_{\text{connected}}(\sum_{\text{pairings}})$, satisfies $\deg \mathcal{P}(X^n) = n - 3$
- ▷ W : Wick map, $W(X^n) = H_n(X; \beta Y)$ is Hermite polynomial (for more general models: Bell polynomial)
- ▷ Θ associated with energy renormalisation, $\Pi_N \Theta(e^{-\alpha X}) = \gamma$

Theorem: [B, Klose, Tapia 25]

$$\log \frac{Z_{N,\alpha}}{Z_{N,0}} = \log \mathbb{E}[e^{-\alpha X - \beta Y}] - \gamma \asymp - \sum_{n \geq 4} \frac{(-\alpha)^n}{n!} \Pi_N \mathcal{A}(\mathcal{P}(X^n))$$

as asymptotic expansion with terms uniformly bounded in N

Main result: Commutative diagram

$$\begin{array}{ccc}
 e^{-\alpha X} & \xrightarrow{\mathcal{P}} & \sum_p \frac{(-\alpha)^p}{p!} \mathcal{P}(X^p) \\
 \downarrow W & & \downarrow M^{\text{BPHZ}+\Theta} \\
 e^{-\alpha X - \beta Y} & \xrightarrow{\mathcal{P}} & \sum_{n,m} \frac{(-\alpha)^m (-\beta)^{n-m}}{m!(n-m)!} \mathcal{P}(X^m Y^{n-m}) \\
 & & \xrightarrow{\Pi_N} \log \mathbb{E}[e^{-\alpha X - \beta Y}]
 \end{array}
 \quad \begin{array}{c}
 \nearrow \Pi_N^{\text{BPHZ}+\Pi_N\Theta} \\
 \searrow
 \end{array}$$

- ▷ $\mathcal{P} = \Pi_{\text{connected}}(\sum_{\text{pairings}})$, satisfies $\deg \mathcal{P}(X^n) = n - 3$
- ▷ W : Wick map, $W(X^n) = H_n(X; \beta Y)$ is Hermite polynomial (for more general models: Bell polynomial)
- ▷ Θ associated with energy renormalisation, $\Pi_N \Theta(e^{-\alpha X}) = \gamma$

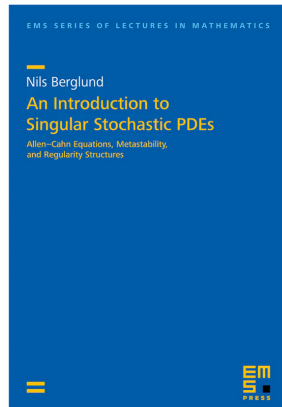
Theorem: [B, Klose, Tapia 25]

$$\log \frac{Z_{N,\alpha}}{Z_{N,0}} = \log \mathbb{E}[e^{-\alpha X - \beta Y}] - \gamma \asymp - \sum_{n \geq 4} \frac{(-\alpha)^n}{n!} \Pi_N \mathcal{A}(\mathcal{P}(X^n))$$

as asymptotic expansion with terms uniformly bounded in N

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- ▷ N. B. & Tom Klose, *Perturbation theory for the Φ_3^4 measure, revisited with Hopf algebras*, arXiv/2207.08555 (2022)
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Thanks for your attention!

Slides available at <https://www.idpoisson.fr/berglund/LJLL26.pdf>

Some literature on the static Φ_3^4 model

- ▷ Glimm & Jaffe (1968, 1973), Feldman (1974):
Combinatorics of Feynman diagrams
- ▷ Benfatto, Cassandro, Gallavotti, Nicolò & Olivieri (1978, 1980):
Renormalisation group (integrating out scales)
- ▷ Brydges, Fröhlich & Sokal (1983):
Generating function and skeleton inequalities
- ▷ Brydges, Dimock & Hurd (1995):
Polymer expansions
- ▷ Connes & Kreimer (2000, 2001):
Hopf algebras
- ▷ ...
- ▷ Barashkov & Gubinelli (2020):
Boué–Dupuis formula