

Yang Mills, unitary Brownian bridge and potential theory under constraint

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- ▶ Unitary Brownian motion : asymptotics
- ▶ Unitary Brownian bridge : shape of the dominant representation
- ▶ Some concluding remarks

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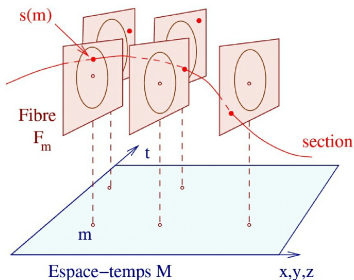
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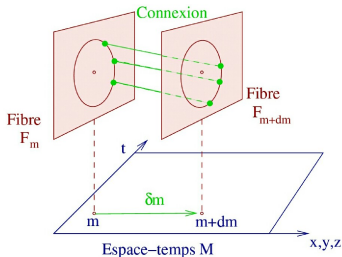
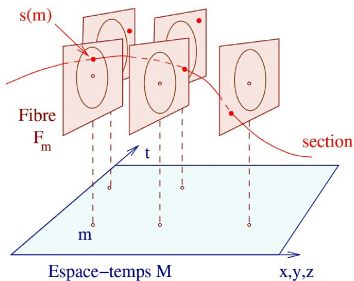
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A lot of results concerning Yang-Mills on a cylinder or a sphere (Douglas-Kazakov, Gross-Matytsin (circa 1995)), in particular

Some properties of large N two-dimensional Yang-Mills theory

[Nucl.Phys. B437 (1995)]

Unitary Brownian motion

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One can define a Brownian motion on the unit circle

$\mathbb{U} := \{z \in \mathbb{C} / |z| = 1\}$, as follows : $U_1(t) = e^{iB(t)}$, where B is a standard Brownian motion on \mathbb{R} .

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For $N \geq 1$, this can be generalized as follows :

$$dU_N(t) = dK_N(t)U_N(t) - \frac{1}{2}U_N(t)dt,$$

with K_N a Brownian motion on $u(N)$ equipped with

$$(X, Y)_{u(N)} = N\text{Tr}(X^* Y).$$

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Poisson summation formula : if $\check{f}(x) = \int_{\mathbb{R}} e^{iux} f(u) du$,

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$$\text{Ex : } p_2 := \sum x_i^2 = \sum_{i \leq j} x_i x_j - \sum_{i < j} x_i x_j = s(2) - s(1,1)$$

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and $\int_{\mathcal{U}(N)} \overline{s_{\alpha}(U)} s_{\beta}(U) dm_N(U) = \delta_{\alpha, \beta} \mathbf{1}_{\ell(\alpha) \leq N}$.

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For any $t > 0$, we denote by ν_t the probability measure on \mathbb{U} such that, for all $n \geq 0$, $\int z^{-n} d\nu_t(z) = \int z^n d\nu_t(z) = c_n(t)$.

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From harmonic analysis, we get that

$$Z_{N,T} = C_{N,T} \sum_{\ell} e^{-N^2 I_T(\hat{\mu}_{\ell})},$$

with

$$I_T(\mu) := - \iint \ln|x-y| d\mu(x) d\mu(y) + \int \frac{T}{2} x^2 d\mu(x)$$

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Proposition

For all $T > 0$,

$$\lim_{N \rightarrow \infty} \frac{1}{N^2} \ln Z_{N,T} = \frac{T}{24} + \frac{3}{2} - \inf I_T(\mu),$$

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Tools : large deviations results.

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- ▶ If $T \leq \pi^2$, the density of μ_T^* with respect to Lebesgue measure is given by

$$\frac{d\mu_T^*(x)}{dx} = \frac{T}{2\pi} \sqrt{\frac{4}{T} - x^2} \mathbf{1}_{\left[-\frac{2}{\sqrt{T}}, \frac{2}{\sqrt{T}}\right]}(x),$$

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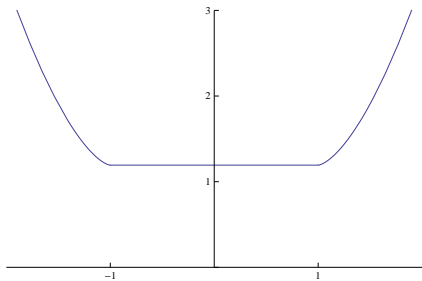
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- ▶ If $T > \pi^2$, the density of μ_T^* is described in terms of elliptic functions.

Consequence : The function F is of class \mathcal{C}^2 on \mathbb{R}_+^* and of class \mathcal{C}^∞ on $\mathbb{R}_+^* \setminus \{\pi^2\}$. At π^2 , $F^{(3)}$ has a discontinuity of first kind.

Potential theory under constraint

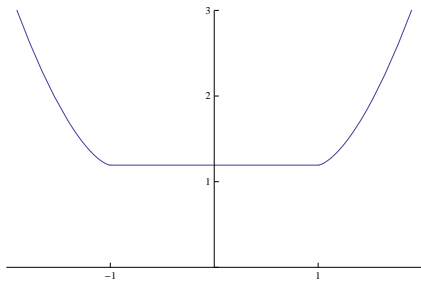
Potential theory under constraint



$$U^\mu + Q \geq C$$

$$U^\mu + Q = C \text{ on the support}$$

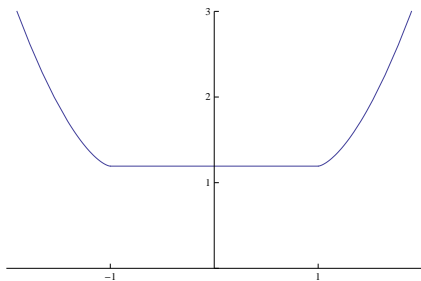
Potential theory under constraint



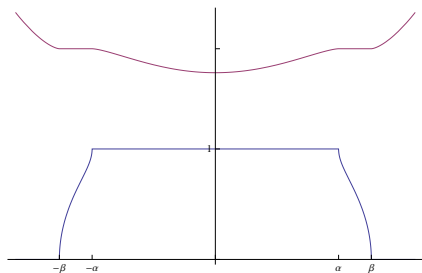
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Potential theory under constraint



$$U^\mu + Q \geq C$$
$$U^\mu + Q = C \text{ on the support}$$



$$U^\mu + Q \geq C \text{ outside the support}$$
$$U^\mu + Q = C \text{ on the "free" part}$$
$$U^\mu + Q \leq C \text{ where it saturates}$$

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- ▶ Fascinating model for which everything can be computed explicitly
- ▶ In a recent work of Liechty and Wang, μ_T^* appears as the equilibrium measure associated to orthogonal polynomials for a discrete gaussian measure (also linked with Unitary brownian bridge)
- ▶ for some parameters (t, T) , the asymptotic spectral measure of uBb is known and related to the family μ_T^* in a way which is still to be understood in details (work in progress with T. Lévy).