

Local Statistics for Particle Systems with Repulsive Interaction

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Asymptotics”

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Topics

- 1 From β -Ensembles to Repulsive Particle Systems.
- 2 How to tackle the new Interaction?
- 3 Arbitrary $\beta > 0$: Bulk Correlations.
- 4 $\beta = 2$:
 - ▶ Correlations under Unfolding.
 - ▶ Empirical Spacings.
 - ▶ Edge Correlations.
 - ▶ Fluctuations and Deviations of the Largest Particle.
- 5 Central Ideas of the Method.

Motivation and Aim

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 - ▶ Spacings of zeros of the Riemann Zeta function on the critical line.
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For β -Ensembles: Show for fixed $\beta > 0$ that local statistics are independent of the potential $Q : \mathbb{R} \rightarrow \mathbb{R}$.

$$P_{N,Q,\beta}(x) := \frac{1}{Z_{N,Q,\beta}} \prod_{i < j} |x_i - x_j|^\beta e^{-N \sum_{j=1}^N Q(x_j)}.$$

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Question: Why not change the interaction potential?

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- Define repulsive particle system on \mathbb{R}^N

$$P_{N,Q,\beta}^h(x) := \frac{1}{Z_{N,Q,\beta}^h} \prod_{i<j} |x_i - x_j|^\beta e^{-\sum_{i<j} h(x_i - x_j)} e^{-N \sum_{j=1}^N Q(x_j)}$$

with additional (smooth) pair potential h .

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- Particles instead of eigenvalues!

How to tackle the new Interaction? Basic Idea I

Assumptions (to be cont.): h symmetric around 0, negative-definite.

Choose Gaussian process $(f(t))_{t \in \mathbb{R}}$ with mean 0 and $\mathbb{E}f(t)f(s) = -h(t - s)$.

Then

$$\sum_{j=1}^N f(x_j) \sim \mathcal{N} \left(0, - \sum_{i,j} h(x_i - x_j) \right) \quad \text{for all } x_1, \dots, x_N \in \mathbb{R} \text{ and}$$

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$$\begin{aligned} \mathbb{E} \exp \left(\sum_{j=1}^N f(x_j) \right) &= \exp \left(- \frac{1}{2} \sum_{i,j} h(x_i - x_j) \right) \\ &= \exp \left(- \sum_{i < j} h(x_i - x_j) - \frac{N}{2} h(0) \right). \end{aligned}$$

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$$\prod_{i < j} |x_i - x_j|^\beta e^{-\sum_{i < j} h(x_i - x_j)} e^{-\sum_{j=1}^N NQ(x_j)} \propto \mathbb{E} \prod_{i < j} |x_i - x_j|^\beta e^{-\sum_{j=1}^N NQ(x_j) + f(x_j)}.$$

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Instead of negative definiteness: Assume h is a Schwartz function.

- Allows for Fourier techniques.
- Can split $h = h^+ - h^-$ with positive-definite functions h^\pm .
- Complex analysis (Vitali's Theorem): Suffices to consider $h_z := zh^+ - h^-$ with negative z .

Repulsive Particles: Summary Assumptions

$$P_{N,Q,\beta}^h(x) = \frac{1}{Z_{N,Q,\beta}^h} \prod_{i < j} |x_i - x_j|^\beta e^{-\sum_{i < j} h(x_i - x_j)} e^{-N \sum_{j=1}^N Q(x_j)}.$$

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- Assumptions on h : symmetric around zero, Schwartz function, real analytic.
- Assumptions on potential Q : symmetric around zero, real analytic and for given h sufficiently uniformly convex:
 $\min_{t \in \mathbb{R}} Q''(t) > C(h) \geq 0$ (ensures uniqueness of equilibrium measure).

Global Correlations

$\rho_{N,Q,\beta}^{h,k}(x_1, \dots, x_k) := \int_{\mathbb{R}^{N-k}} P_{N,Q,\beta}^h(x) dx_{k+1} \dots dx_N$: k -th correlation function.

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Theorem (Götze-V., AoP. '14)

There exists $\mu_{Q,\beta}^h$ probability measure with compact connected support and positive density on the interior, s.th.

$$\rho_{N,Q,\beta}^{h,k}(dt) \rightarrow (\mu_{Q,\beta}^h)^{\otimes k} \text{ weakly, as } N \rightarrow \infty.$$

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- Global correlations first announced in Boutet de Monvel, Pastur, Shcherbina'95 for convex many-body interactions, expansion of the partition function in Borot, Guionnet, Kozłowski'15.
- Somewhat related interactions occur for multi-matrix models (Figalli, Guionnet'14).

Local Bulk Correlations for arbitrary β : Averaged Vague Convergence

Compare local correlations of $P_{N,Q,\beta}^h$ with those of the Gaussian β -Ensemble $P_{N,G,\beta}$, i.e. $G(t) := t^2, \mu_{G,\beta}$ semicircle law, $\rho_{N,G,\beta}^k$ k -th correlation function.

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Theorem (V., ECP '13)

Let $0 < \xi \leq 1/2$ and $s_N := N^{-1+\xi}$.

For $k = 1, 2, \dots$, any $a \in \text{supp}(\mu_{Q,\beta}^h)^\circ$, any $a' \in \text{supp}(\mu_{G,\beta})^\circ$, any smooth function $f : \mathbb{R}^k \rightarrow \mathbb{R}$ with compact support

$$\begin{aligned} & \lim_{N \rightarrow \infty} \int dt^k f(t) \\ & \int_{a-s_N}^{a+s_N} \frac{1}{\mu_{Q,\beta}^h(a)^k} \rho_{N,Q,\beta}^{h,k} \left(u + \frac{t_1}{N\mu_{Q,\beta}^h(a)}, \dots, u + \frac{t_k}{N\mu_{Q,\beta}^h(a)} \right) \frac{du}{2s_N} \\ & - \int_{a'-s_N}^{a'+s_N} \frac{1}{\mu_{G,\beta}(a')^k} \rho_{N,G,\beta}^k \left(u' + \frac{t_1}{N\mu_{G,\beta}(a')}, \dots, u' + \frac{t_k}{N\mu_{G,\beta}(a')} \right) \frac{du'}{2s_N} \\ & = 0. \end{aligned}$$

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For $k \geq 1$ we have uniformly on compacts in t_1, \dots, t_k and uniformly in $a \in I$, $I \subset (\text{supp } \mu_Q^h)^\circ$ compact

$$\begin{aligned} \lim_{N \rightarrow \infty} \frac{1}{\mu_Q^h(a)^k} \rho_{N,Q}^{h,k} \left(a + \frac{t_1}{N\mu_Q^h(a)}, \dots, a + \frac{t_k}{N\mu_Q^h(a)} \right) \\ = \det \left[\frac{\sin(\pi(t_i - t_j))}{\pi(t_i - t_j)} \right]_{1 \leq i, j \leq k}. \end{aligned}$$

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Analogous results for β -Ensembles:

- $\beta = 2$: Deift et al.'99; Pastur, Shcherbina'97,'08; Levin, Lubinsky'08–, many more.
- General β : Valko, Virag'09; Bourgade, Erdős, Yau'14; Shcherbina'14; Bekerman, Figalli, Guionnet'15. Compare with Figalli, Guionnet'14.

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uniformly for $t_1, \dots, t_k = o(N)$.

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- Problem is that $\mu_Q^h(a)$ is not constant.

Local Bulk Correlations for $\beta = 2$: Unfolding

- Unfolding: $NF_Q^h(x_1), \dots, NF_Q^h(x_N)$ with F_Q^h distribution function of μ_Q^h .
Works throughout the whole spectrum (excluding the edges).

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With $I_N \subset [0, N]$ such that $\frac{1}{N} \text{dist}(I_N, \{0, N\}) \geq c > 0$ for N large enough

Theorem (Schubert-V., EJP '15)

With $\widehat{t}_j := (F_Q^h)^{-1}(t_j/N)$

$$\frac{1}{\prod_{j=1}^k \mu_Q^h(\widehat{t}_j)} \rho_{N,Q}^{h,k}(\widehat{t}_1, \dots, \widehat{t}_k) = \det \left[\frac{\sin(\pi(t_i - t_j))}{\pi(t_i - t_j)} \right]_{1 \leq i, j \leq k} + o(1)$$

uniform for $t_1, \dots, t_k \in I_N$.

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- Transform μ_Q^h to uniform distribution.
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uniform for $t_1, \dots, t_k \in I_N$ and any $\varepsilon > 0$.

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Theorem (Schubert-V., EJP '15)

With $\widehat{t}_j := (F_Q^h)^{-1}(t_j/N)$ and $h = 0$ ($\beta = 2$ -Ensemble),

$$\frac{1}{\prod_{j=1}^k \mu_Q(\widehat{t}_j)} \rho_{N,Q}^k(\widehat{t}_1, \dots, \widehat{t}_k) = \det \left[\frac{\sin(\pi(t_i - t_j))}{\pi(t_i - t_j)} \right]_{1 \leq i, j \leq k} + \mathcal{O}\left(\frac{1}{N}\right)$$

uniform for $t_1, \dots, t_k \in I_N$ and any $\varepsilon > 0$.

Nearest Neighbor Spacings

- Let $x_1 \leq x_2 \leq \dots \leq x_N$ and I_N interval, nearest neighbor spacings in I_N :

$$\sigma(I_N, x) := \sum_{x_j, x_{j+1} \in I_N} \delta_{x_{j+1} - x_j}.$$

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- Classical result (Deift et al. 99): $a \in \text{supp} \mu_Q^\circ$, $t_N \rightarrow \infty$ and $t_N/N \rightarrow 0$, s given, localized scaling $x_{\text{loc}} := N\mu_Q(a)x$,

$$\lim_{N \rightarrow \infty} \mathbb{E}_{N,Q} \frac{1}{2t_N \mu_Q(a)} \int_0^s d\sigma((a - t_N, a + t_N), x_{\text{loc}}) = G(s),$$

Nearest Neighbor Spacings

- Let $x_1 \leq x_2 \leq \dots \leq x_N$ and I_N interval, nearest neighbor spacings in I_N :

$$\sigma(I_N, x) := \sum_{x_j, x_{j+1} \in I_N} \delta_{x_{j+1} - x_j}.$$

- Classical result (Deift et al. 99): $a \in \text{supp} \mu_Q^\circ$, $t_N \rightarrow \infty$ and $t_N/N \rightarrow 0$, s given, localized scaling $x_{\text{loc}} := N\mu_Q(a)x$,

$$\lim_{N \rightarrow \infty} \mathbb{E}_{N,Q} \frac{1}{2t_N \mu_Q(a)} \int_0^s d\sigma((a - t_N, a + t_N), x_{\text{loc}}) = G(s),$$

with G being the distribution function of the Gaudin distribution,

$$G(s) := \sum_{k \geq 2} \frac{(-1)^k}{(k-1)!} \int_{[0,s]^{k-1}} \det[(S(z_i - z_j))_{1 \leq i, j \leq k}]|_{z_1=0} dz_2 \dots dz_k,$$

where S is the sine kernel.

Nearest Neighbor Spacings: Kolmogorov Distance

- More recently (Tao'13; Erdős, Yau'15; Bekerman, Figalli, Guionnet'15; cf. Figalli, Guionnet'14) for any index set I excluding edge indices

$$\lim_{N \rightarrow \infty} \mathbb{E}_{N,Q} \frac{1}{|I|} \sum_{i \in I} g(\mu_Q(q_i) N(x_{i+1} - x_i)) = \int g(s) dG(s)$$

with q_i being the i/N -quantile of μ_Q , g test function.

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$$\lim_{N \rightarrow \infty} \mathbb{E}_{N,Q} \left(\sup_{s \in \mathbb{R}} \left| \int_0^s \frac{1}{2t_N \mu_Q(a)} d\sigma((a - t_N, a + t_N), x_{\text{loc}}) - G(s) \right| \right) = 0.$$

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However, only a tiny fraction of all spacings is considered!

Nearest Neighbor Spacings: New Results

$$\widehat{\sigma}(I_N, \mathbf{x}) := \frac{1}{\int_0^\infty d\sigma(I_N, \mathbf{x})} \sigma(I_N, \mathbf{x}), \quad (\mathbf{x}_{\text{unf}})_i := NF_Q^h(x_i).$$

Theorem (Schubert-V., EJP'15)

$$\lim_{N \rightarrow \infty} \mathbb{E}_{N, Q}^h \left(\sup_{s \in \mathbb{R}} \left| \int_0^s d\widehat{\sigma}([0, N], \mathbf{x}_{\text{unf}}) - G(s) \right| \right) = 0.$$

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Let h **negative-definite**. If $I_N \subset [0, N]$ with $\frac{1}{N} \text{dist}(I_N, \{0, N\}) \geq c > 0$ for N large enough, then for any $\varepsilon > 0$

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If h **arbitrary**, then error is $o(1)$.

Edge Correlations for Repulsive Particles

$$P_{N,Q}^h(x) = \frac{1}{Z_{N,Q}^h} \prod_{i<j} |x_i - x_j|^2 e^{-\sum_{i<j} h(x_i - x_j) - N \sum_{j=1}^N Q(x_j)}.$$

Theorem (Kriecherbauer-V., '15)

Let $\text{supp} \mu_Q^h = [-b, b]$. There exists $c^* > 0$ such that for $q < 0 < p$ and $t \in [q, pN^{4/15}]^k$

$$\left(\frac{N^{1/3}}{c^*}\right)^k \rho_{N,Q}^{h,k} \left(b + \frac{t_1}{c^* N^{2/3}}, \dots, b + \frac{t_k}{c^* N^{2/3}}\right) = \det [K_{Ai}(t_i, t_j)]_{1 \leq i, j \leq k} + o(1).$$

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Edge universality for β -Ensembles: Ramirez, Rider, Virag'11; Krishnapur, Rider, Virag'15;

Bourgade, Erdős, Yau'14; Bekerman, Figalli, Guionnet'15; cf. Figalli, Guionnet'14

Asymptotics for the largest Particle: Fluctuations and Deviations

Theorem (Kriecherbauer-V., '15)

For any (fixed) $s \in \mathbb{R}$,

$$P_{N,Q}^h \left((x_{\max} - b) c^* N^{2/3} \leq s \right) = F_{\text{TW}}(s) + o(1).$$

where F_{TW} is the distribution function of the ($\beta = 2$) Tracy-Widom distribution.

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- Large deviations: $s \rightarrow \infty$ and $t = b + \frac{s}{c^* N^{2/3}} > b$. Non-universal!

Moderate Deviations: Logarithmic Form

Moderate deviations in logarithmic form:

Theorem (Kriecherbauer-V., '15)

$$\frac{\log P_{N,Q}^h((x_{\max} - b)c^* N^{2/3} > s)}{s^{3/2}} = -\frac{4}{3} + o(1)$$

with $o(1)$ uniform in $s \in [1, o(N^{2/3})]$.

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with $o(1)$ uniform in $s \in [1, o(N^{2/3})]$. For h negative definite, replace $o(1)$ by $-\frac{\log(16\pi s^{3/2})}{s^{3/2}} + \mathcal{O}\left(\frac{s}{N^{2/3}}\right) + \mathcal{O}\left(\frac{1}{s^3}\right)$.

Uses known right-tail asymptotics for Tracy-Widom distribution:

$$1 - F_{\text{TW}}(s) = \frac{1}{16\pi} \frac{e^{-\frac{4}{3}s^{3/2}}}{s^{3/2}} \left(1 + \mathcal{O}\left(\frac{1}{s^{3/2}}\right)\right), \quad \text{for } s \rightarrow \infty.$$

Moderate Deviations: Non-Logarithmic Form

Theorem (Kriecherbauer-V., '15)

$$\frac{P_{N,Q}^h((x_{\max} - b)c^* N^{2/3} > s)}{(1 - F_{TW}(s))} = \exp\left(s^{3/2} \sum_{j=1}^{\infty} d_j \left(\frac{s}{N^{2/3}}\right)^j\right) (1 + o(1))$$

uniformly for $s = o(N^{2/3})$ with coefficients d_j depending on Q and h .

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- To be compared with result in classical probability (Cramér'38) for sums of i.i.d. r.v. X_i with mean μ and variance σ^2 :

$$\frac{P\left(\frac{\sum_{i=1}^N X_i - \mu}{\sigma\sqrt{N}} > s\right)}{(1 - \Phi(s))} = \exp\left(s^2 \sum_{j=1}^{\infty} a_j \left(\frac{s}{\sqrt{N}}\right)^j\right) (1 + o(1)).$$

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- Describes the transition from universality to non-universality!

Asymptotics for the largest Particle: Large Deviations

Theorem (Kriecherbauer-V., '15)

Let $M > b$ and $0 < \varepsilon < 2/3$. Then uniformly for $t \in [b + N^{-\varepsilon}, M]$ (large deviations),

$$\frac{\log P_{N,Q}^h(x_{\max} > t)}{N} = -\eta_{Q,h}(t) - \frac{\log(N(t-b)^{3/2})}{N} + \mathcal{O}\left(\frac{1}{N}\right),$$

where the \mathcal{O} term is uniform in N and in $t \in (b + N^{-2/3}, T)$ and the rate function is

$$\eta_{Q,h}(t) = -2 \int \log|t-s| d\mu_Q^h(s) + \int h(t-s) d\mu_Q^h(s) + Q(t) - C.$$

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- Without rate in Borot,Guionnet,Kozłowski'15. Non-logarithmic form for $\beta = 2$ -Ensembles in Eichelsbacher,Kriecherbauer,Schüler'16.
- Conjecture new source of non-universality.

Ideas of the Method: Determining the Equilibrium Measure 1

$$P_{N,Q,\beta}^h(x) = \frac{1}{Z_{N,Q,\beta}^h} \prod_{i<j} |x_j - x_i|^\beta e^{-\sum_{i<j} h(x_i - x_j) - N \sum_{j=1}^N Q(x_j)}.$$

- Idea: Compare $P_{N,Q,\beta}^h$ with β -Ensemble $P_{N,V,\beta}$ that has the same limiting measure.

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- Idea: Compare $P_{N,Q,\beta}^h$ with β -Ensemble $P_{N,V,\beta}$ that has the same limiting measure.
- Let μ probability measure on \mathbb{R} . Hoeffding type decomposition

$$\sum_{i<j} h(x_i - x_j) = c_N + N \sum_{j=1}^N h_\mu(x_j) - \mathcal{U}_\mu(x),$$

where c_N is a constant, $h_\mu(t) := \int h(t - s) d\mu(s)$ and \mathcal{U}_μ is the quadratic statistic

$$\mathcal{U}_\mu(x) := -\frac{1}{2} \left(\sum_{i,j=1}^N h(x_i - x_j) - h_\mu(x_i) - h_\mu(x_j) + h_{\mu\mu} \right).$$

Ideas of the Method: Determining the Equilibrium Measure 2

Now, with $V_\mu(t) := Q(t) + h_\mu(t)$

$$\begin{aligned} P_{N,Q,\beta}^h(x) &= \frac{1}{Z_{N,Q,\beta}^h} \prod_{i<j} |x_i - x_j|^\beta e^{-\sum_{i<j} h(x_i - x_j) - N \sum_{j=1}^N Q(x_j)} \\ &= \frac{e^{-c_N} Z_{N,V_\mu,\beta}}{Z_{N,Q,\beta}^h} \frac{1}{Z_{N,V_\mu,\beta}} \prod_{i<j} |x_i - x_j|^\beta e^{-N \sum_{j=1}^N V_\mu(x_j) + \mathcal{U}_\mu(x)} \end{aligned}$$

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- Choose μ such that \mathcal{U}_μ is small under $P_{N,V_\mu,\beta}$. But \mathcal{U}_μ is centered under μ ! Thus

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- Choose μ such that \mathcal{U}_μ is small under $P_{N,V_\mu,\beta}$. But \mathcal{U}_μ is centered under μ ! Thus μ must be chosen as the equilibrium measure to V_μ !

Ideas of the Method: Determining the Equilibrium Measure 2

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$$\begin{aligned} P_{N,Q,\beta}^h(x) &= \frac{1}{Z_{N,Q,\beta}^h} \prod_{i<j} |x_i - x_j|^\beta e^{-\sum_{i<j} h(x_i - x_j) - N \sum_{j=1}^N Q(x_j)} \\ &= \frac{e^{-c_N} Z_{N,V_\mu,\beta}}{Z_{N,Q,\beta}^h} \frac{1}{Z_{N,V_\mu,\beta}} \prod_{i<j} |x_i - x_j|^\beta e^{-N \sum_{j=1}^N V_\mu(x_j) + \mathcal{U}_\mu(x)} \end{aligned}$$

- Choose μ such that \mathcal{U}_μ is small under $P_{N,V_\mu,\beta}$. But \mathcal{U}_μ is centered under μ ! Thus μ must be chosen as the equilibrium measure to V_μ !
- Solve this implicit problem by applying a fixed point theorem: get $\mu_{Q,\beta}^h$.

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 $V := Q + h_{\mu_{Q,\beta}^h}$.
- Show that $\mathcal{U} = \mathcal{U}_{\mu_{Q,\beta}^h}$ is indeed just a perturbation. Here convexity used (concentration inequalities).

Ideas of the Method: Linearization

- Linearization trick: Recall that for $-h$ positive definite ($-\hat{h} \geq 0$)

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$$\exp(\mathcal{U}(x)) = \mathbb{E} \exp\left(\sum_{j=1}^N f(x_j) - N \int f d\mu_{Q,\beta}^h\right).$$

- Now add f to the potential: ensembles with densities proportional to

$$\prod_{i < j} |x_i - x_j|^\beta e^{-\sum_{j=1}^N N V(x_j) + f(x_j)}.$$

Perturbation f does not change global or local asymptotics.

Summary

$$P_{N,Q,\beta}^h(x) = \frac{1}{Z_{N,Q,\beta}^h} \prod_{i < j} |x_i - x_j|^\beta e^{-\sum_{i < j} h(x_i - x_j) - N \sum_{j=1}^N Q(x_j)}$$

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Thank you for your attention!