

$\overline{\partial}$ Problems in Random Matrix Theory

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Abstract

A $\overline{\partial}$ problem is a kind of generalization of a Riemann-Hilbert problem. We will describe how some $\overline{\partial}$ problems arise in the context of the orthogonal polynomial approach to random matrix theory.



Circular Ensembles and Universality

Consider the group $U(N)$ of $N \times N$ unitary matrices \mathbf{U} equipped with a probability measure of the form

$$dp(\mathbf{U}) = \frac{1}{Z_N} e^{-N \text{Tr} V(\arg(\mathbf{U}))} dH_N(\mathbf{U}),$$

where dH_N denotes Haar measure and Z_N is a normalization constant (partition function). This is the circular ensemble with weight e^{-NV} .

Spectral theorem: diagonalize \mathbf{U} and integrate out the eigenvector variables (Haar measure again). What remains is the joint law for the eigenvalues $\{z_n = e^{i\theta_n}\}_{n=1}^N$:

$$dp(\theta_1, \dots, \theta_N) = \frac{1}{Z'_N} \cdot \prod_{m < n} |e^{i\theta_m} - e^{i\theta_n}|^2 \cdot \prod_{n=1}^N e^{-NV(\theta_n)} d\theta_n.$$

Circular Ensembles and Universality

Correlation functions are expressed in terms of orthogonal polynomials:

$$R_N^{(n)}(\theta_1, \dots, \theta_n) = \det(K_N(\theta_j, \theta_k))_{j,k=1,\dots,n},$$

where

$$K_N(\theta, \phi) = \sum_{n=0}^{N-1} p_n(e^{i\theta}) \overline{p_n(e^{i\phi})} e^{-NV(\theta)/2} e^{-NV(\phi)/2},$$

and $p_n(z) = \gamma_{n,n} z^n + \dots$ satisfies

$$\frac{1}{2\pi} \int_{-\pi}^{\pi} p_n(e^{i\theta}) \overline{p_m(e^{i\theta})} e^{-NV(\theta)} d\theta = \delta_{n,m}.$$

Christoffel-Darboux formula:

$$K_N(\theta, \phi) = \frac{p_N^*(e^{i\theta}) \overline{p_N^*(e^{i\phi})} - p_N(e^{i\theta}) \overline{p_N(e^{i\phi})}}{1 - e^{i(\theta-\phi)}},$$

where if $p(z) = c_0 + c_1 z + \dots + c_n z^n$, then $p^*(z) := \overline{c_0} z^n + \overline{c_1} z^{n-1} + \dots + \overline{c_n}$.

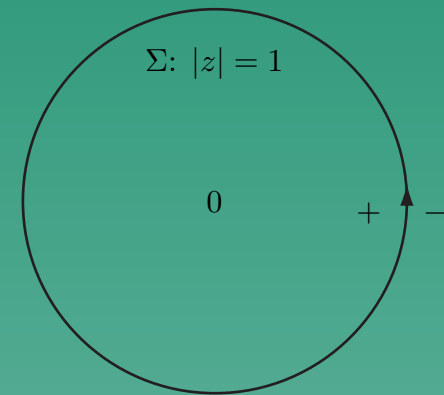
Circle Polynomials: Riemann-Hilbert Problem

Let $e^{-V(\theta)}$ be a weight on the unit circle. Seek $\mathbf{M}^n(z)$, a 2×2 matrix, with the following properties:

Analyticity: $\mathbf{M}^n(z)$ is analytic for $z \in \mathbb{C} \setminus \Sigma$, taking continuous boundary values $\mathbf{M}^n_{\pm}(z)$ for $|z| = 1$.

Jump Condition: The boundary values are related by

$$\mathbf{M}^n_+(e^{i\theta}) = \mathbf{M}^n_-(e^{i\theta}) \begin{bmatrix} 1 & e^{-V(\theta)} e^{-in\theta} \\ 0 & 1 \end{bmatrix}.$$



Normalization: The matrix $\mathbf{M}^n(z)$ satisfies

$$\lim_{z \rightarrow \infty} \mathbf{M}^n(z) \begin{bmatrix} z^{-n} & 0 \\ 0 & z^n \end{bmatrix} = \mathbb{I}.$$

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3. Because $M_{12+}^n(e^{i\theta}) = M_{11}^n(e^{i\theta})e^{-NV(\theta)}e^{-in\theta} + M_{12-}^n(e^{i\theta})$, by the Plemelj formula,

$$M_{12}^n(z) = \frac{1}{2\pi i} \oint \frac{M_{11}^n(s)e^{-NV(\arg(s))}s^{-n}}{s-z} ds + e(z)$$

where $e(z)$ is an entire function.

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4. Because $M_{12}^n(z)z^n \rightarrow 0$ as $z \rightarrow \infty$, we must have $e(z) \equiv 0$ and

$$\frac{1}{2\pi} \int_{-\pi}^{\pi} M_{11}^n(e^{i\theta})e^{-im\theta}e^{-NV(\theta)} d\theta = 0, \quad m = 0, 1, 2, \dots, n-1.$$

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This result identifies $M_{11}^n(z)$ with $\pi_n(z)$, the monic orthogonal polynomial of degree n .

Analytic Weights: Steepest Descent Asymptotics

The simplest case is to take $N = 1$ and let $n \rightarrow \infty$. Make the substitution

$$\mathbf{N}^n(z) := \begin{cases} \mathbf{M}^n(z), & |z| < 1, \\ \mathbf{M}^n(z)z^{-n\sigma_3}, & |z| > 1. \end{cases}$$

This removes the non-identity asymptotics for large z and the jump condition for $\mathbf{N}^n(z)$ becomes:

$$\mathbf{N}_+^n(e^{i\theta}) = \mathbf{N}_-^n(e^{i\theta}) \begin{bmatrix} e^{in\theta} & e^{-V(\theta)} \\ 0 & e^{-in\theta} \end{bmatrix}, \quad z \in \Sigma.$$

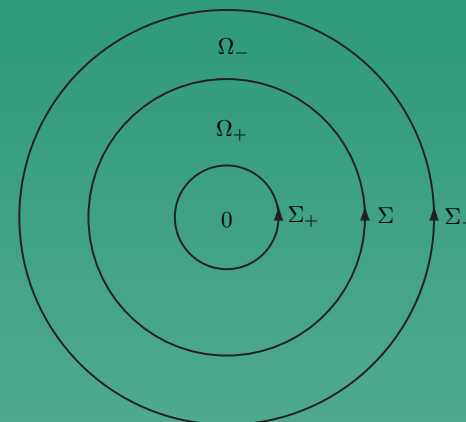
Then, note the factorization:

$$\begin{bmatrix} e^{in\theta} & e^{-V(\theta)} \\ 0 & e^{-in\theta} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ e^{-in\theta}e^{V(\theta)} & 1 \end{bmatrix} \begin{bmatrix} 0 & e^{-V(\theta)} \\ -e^{V(\theta)} & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ e^{in\theta}e^{V(\theta)} & 1 \end{bmatrix}.$$

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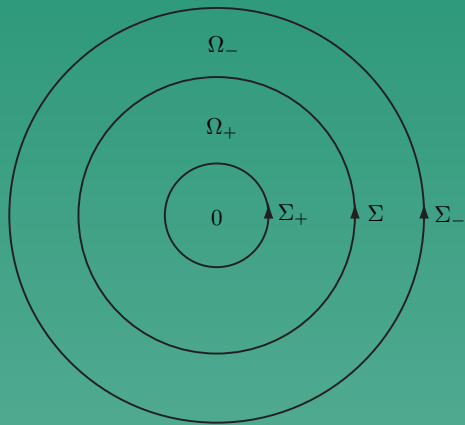
When V is analytic, we may create a new piecewise-analytic unknown as follows:

$$\mathbf{P}^n(z) := \begin{cases} \mathbf{N}^n(z) \begin{bmatrix} 1 & 0 \\ -z^n e^{V(-i \log(z))} & 1 \end{bmatrix}, & z \in \Omega_+, \\ \mathbf{N}^n(z) \begin{bmatrix} 1 & 0 \\ z^{-n} e^{V(-i \log(z))} & 1 \end{bmatrix}, & z \in \Omega_-, \\ \mathbf{N}^n(z), & \text{otherwise.} \end{cases}$$



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Then the jump condition for $\mathbf{P}^n(z)$ is exponentially negligible for large n except:

$$\mathbf{P}_+(e^{i\theta}) = \mathbf{P}_-(e^{i\theta}) \begin{bmatrix} 0 & e^{-V(\theta)} \\ -e^{V(\theta)} & 0 \end{bmatrix}, \quad z = e^{i\theta} \in \Sigma.$$

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This problem has a unique explicit solution in terms of the Szegő function $S(z)$:

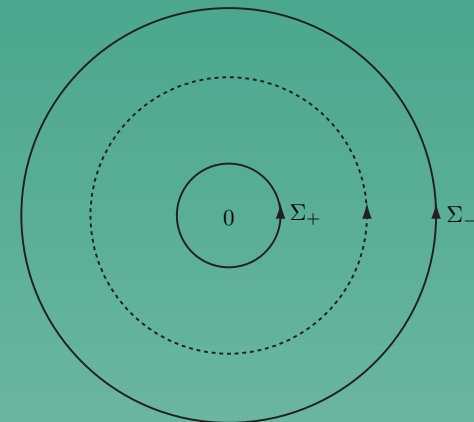
$$\dot{\mathbf{P}}(z) := \begin{cases} \begin{bmatrix} S(z) & 0 \\ 0 & S(z)^{-1} \end{bmatrix}, & |z| > 1 \\ \begin{bmatrix} 0 & S(z) \\ -S(z)^{-1} & 0 \end{bmatrix}, & |z| < 1, \end{cases} \quad S(z) := \exp \left(-\frac{1}{2\pi i} \oint_{\Sigma} \frac{V(\arg(s)) ds}{s - z} \right).$$

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It remains to control the errors. Compare $\mathbf{P}(z)$ with $\dot{\mathbf{P}}(z)$. Define the discrepancy: $\mathbf{H}^n(z) := \mathbf{P}(z)\dot{\mathbf{P}}(z)^{-1}$. This matrix is analytic except on Σ_{\pm} .



Analytic Weights: Steepest Descent Asymptotics

The discrepancy matrix $\mathbf{H}^n(z)$ satisfies a “small-norm” Riemann-Hilbert problem for n large: seek $\mathbf{H}^n(z)$, 2×2 , with the following properties:

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Riemann-Hilbert problems are equivalent to systems of singular integral equations (Cauchy kernels) on the system of jump contours. The integral equations for small-norm problems can be solved by Neumann series. This yields: $\mathbf{H}^n(z) \approx \mathbb{I}$, with error terms given by an asymptotic series.

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Let $x = r \cos \theta$ and $y = r \sin \theta$ where $z = x + iy$. Here is a formula for an “almost analytic extension” of $V(\theta)$:

$$E_m V(r, \theta) := \sum_{p=0}^{m-1} \frac{V^{(p)}(\theta)}{p!} (-i \log(r))^p .$$

Note that Σ is characterized by $r = 1$, or equivalently $\log(r) = 0$. Therefore $E_m V(1, \theta) = V(\theta)$ so we have indeed defined an extension of V from the unit circle.

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What about “near analyticity”? Analytic functions f are characterized by the Cauchy-Riemann equations $\bar{\partial}f = 0$ where

$$\bar{\partial} := \frac{\partial}{\partial \bar{z}} = \frac{1}{2} \left(\frac{\partial}{\partial x} + i \frac{\partial}{\partial y} \right) = \frac{e^{i\theta}}{2} \left(\frac{\partial}{\partial r} + \frac{i}{r} \frac{\partial}{\partial \theta} \right).$$

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Applying $\bar{\partial}$ to $E_m V(r, \theta)$:

$$\bar{\partial} E_m V(r, \theta) = \frac{ie^{i\theta}}{2r(m-1)!} V^{(m)}(\theta) (-i \log(r))^{m-1} \quad (\text{sum telescopes}).$$

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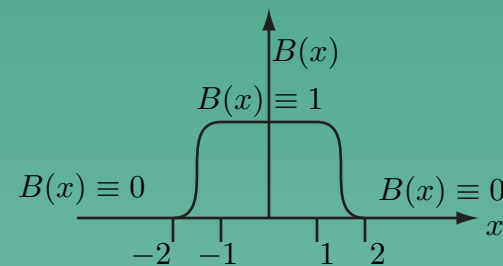
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(If V is analytic, then the infinite series $E_\infty V(r, \theta)$ converges uniformly for r in a neighborhood of $r = 1$ and $\bar{\partial} E_\infty V(r, \theta) = 0$; in other words, $E_\infty V(r, \theta)$ is a series representation of the analytic continuation of V .)

Nonanalytic Weights: $\bar{\partial}$ Steepest Descent Method

We use an extension $E_m V$ of V to make use of the factorization of the jump matrix for $\mathbf{N}^n(z)$:

$$\mathbf{P}_m^n(r, \theta) := \begin{cases} \mathbf{N}^n(re^{i\theta}) \begin{bmatrix} 1 & 0 \\ (re^{i\theta})^{-n} B(\log(r)) e^{E_m V(r, \theta)} & 1 \end{bmatrix}, & r > 1 \\ \mathbf{N}^n(re^{i\theta}) \begin{bmatrix} 1 & 0 \\ -(re^{i\theta})^n B(\log(r)) e^{E_m V(r, \theta)} & 1 \end{bmatrix}, & r < 1. \end{cases}$$



Here $B(\cdot)$ is a C^∞ “bump function”:

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We may expect that $\mathbf{P}_m^n(r, \theta) \approx \dot{\mathbf{P}}(z)$ in some sense. To analyze, define the discrepancy by $\mathbf{H}_m^n(r, \theta) := \mathbf{P}_m^n(r, \theta) \dot{\mathbf{P}}(re^{i\theta})^{-1}$.

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The discrepancy $\mathbf{H}_m^n(r, \theta)$ satisfies another kind of problem, a $\bar{\partial}$ problem: seek $\mathbf{H}_m^n(r, \theta)$, 2×2 , with the following properties:

Smoothness: $\mathbf{H}_m^n(r, \theta)$ is a Lipschitz continuous function on the whole polar plane.

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where $\mathbf{W}_m^n(r, \theta)$ is known.

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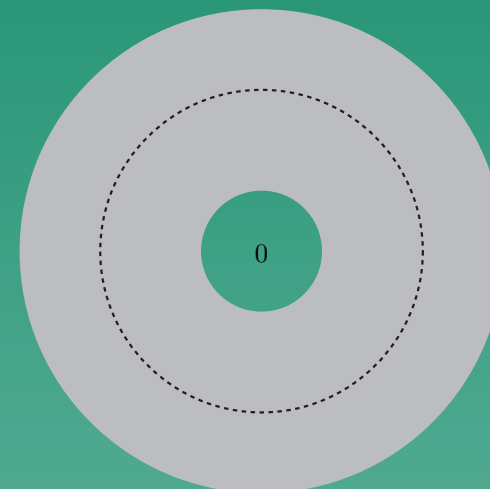
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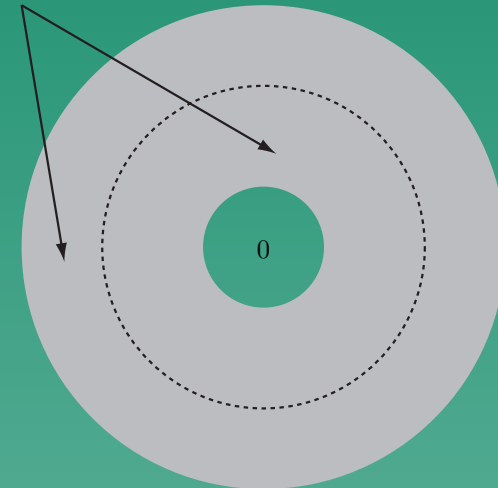
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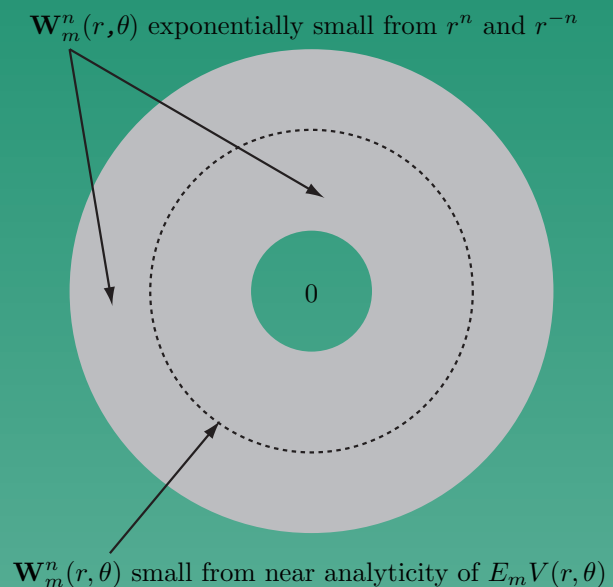
The matrix $\mathbf{W}_m^n(r, \theta)$ is nonzero only in the annulus $|\log(r)| < 2$. Moreover, it is small when n is large:

$\mathbf{W}_m^n(r, \theta)$ exponentially small from r^n and r^{-n}



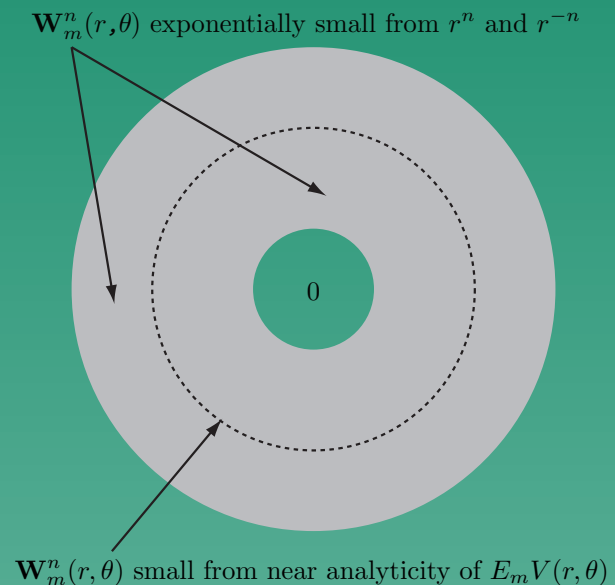
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This makes the $\bar{\partial}$ problem for $\mathbf{H}_m^n(r, \theta)$ a kind of small-norm problem that can be analyzed with great precision, more easily than small-norm Riemann-Hilbert problems due to local integrability of the Cauchy kernel on the plane.

Nongaussian Unitary Ensemble With Convex Exponential Weights

In a similar way as for the circular ensembles, the measure on $N \times N$ Hermitian matrices

$$dp(\mathbf{M}) = \frac{1}{Z_N} e^{-N\text{tr}(V(\mathbf{M}))} d\mathbf{M}, \quad d\mathbf{M} = \text{Lebesgue measure on independent entries}$$

leads to the joint law for the real eigenvalues $x_1 \leq \dots \leq x_N$:

$$dp(x_1, \dots, x_N) = \frac{1}{Z'_N} \cdot \prod_{m < n} (x_n - x_m)^2 \cdot \prod_{n=1}^N e^{-NV(x_n)} dx_n.$$

The correlation functions have determinantal form

$$R_N^{(n)}(x_1, \dots, x_n) = \det(K_N(x_j, x_k))_{j,k=1,\dots,n} \text{ with kernel}$$

$$K_N(x, y) := \sum_{n=0}^{N-1} p_n(x)p_n(y)e^{-NV(x)/2}e^{-NV(y)/2},$$

and $p_n(x)$ is the orthonormal polynomial of degree n for the measure $e^{-NV(x)} dx$ on \mathbb{R} .

Orthogonal Polynomials on the Real Line: Nonanalytic Weights

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In work in progress with K. McLaughlin, we are extending the $\bar{\partial}$ method to handle support endpoints. Our aim is to establish universality of key limiting kernels (sine kernel in the bulk, Airy kernel at the edge leading to the Tracy-Widom law for the fluctuations of the extreme eigenvalues) describing local eigenvalue statistics, *beyond the analytic class of weights V* .

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Consider the set \mathcal{N}_N of $N \times N$ complex matrices \mathbf{M} that are *normal*: $[\mathbf{M}, \mathbf{M}^\dagger] = 0$. Let $\mathbf{X} := \frac{1}{2}(\mathbf{M} + \mathbf{M}^\dagger)$ and $\mathbf{Y} := \frac{1}{2i}(\mathbf{M} - \mathbf{M}^\dagger)$ be the Hermitian “real” and “imaginary” parts of \mathbf{M} , and let $V(x, y)$ be a real function on \mathbb{R}^2 with sufficient growth at infinity. Equip \mathcal{N}_N with the probability distribution

$$dp(\mathbf{M}) = \frac{1}{Z_N} e^{-N \text{tr}(V(\mathbf{X}, \mathbf{Y}))} d\mu(\mathbf{M})$$

where μ is the measure on \mathcal{N}_N induced by the flat metric on all $N \times N$ complex matrices.

Normal Matrix Models

Diagonalization of \mathbf{M} and integrating out the eigenvectors yields the joint law for the (generally complex) eigenvalues $z_n = x_n + iy_n$ in the form

$$dp(x_1, y_1, \dots, x_N, y_N) = \frac{1}{Z'_N} \prod_{j < k} |z_j - z_k|^2 \cdot \prod_{n=1}^N e^{-NV(x_n, y_n)} dx_n dy_n.$$

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There is, unfortunately, no Christoffel-Darboux formula to telescope the sum in $K_N(z, w)$. Nonetheless, information about the asymptotic behavior of eigenvalue statistics lies in the large degree behavior of these orthogonal polynomials.

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The orthogonal polynomials with respect to $e^{-NV(x,y)} dx dy$ are characterized directly by a slightly modified $\bar{\partial}$ problem. Let n and N be positive integers, with $n \leq N$, and seek $\mathbf{M}^n(x, y)$, a 2×2 matrix-valued function on \mathbb{R}^2 with the following properties:

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Then, $M_{11}(x, y)$ is the monic orthogonal polynomial of degree n in $z = x + iy$, with respect to the measure $e^{-NV(x,y)} dx dy$.

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- However, a genuinely two-dimensional analogue of the three-factor factorization and subsequent deformation of Riemann-Hilbert problems is required for this problem. This is the subject of current work.

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Thank You!