



The Centrally Symmetric V -States for Active Scalar Equations. Two-Dimensional Euler with Cut-Off

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Abstract: We consider the family of active scalar equations on the plane and study the dynamics of two centrally symmetric patches. We focus on the two-dimensional Euler equation written in the vorticity form and consider its truncated version. For this model, a non-linear and non-local evolution equation is studied and a family of stationary solutions $\{y(x, \lambda)\}$, $x \in [-1, 1]$, $\lambda \in (0, \lambda_0)$ is found. For these functions, we have $y(x, \lambda) \in C^\infty(-1, 1)$ and $\|y(x, \lambda) - |x|\|_{C[-1,1]} \rightarrow 0$, $\lambda \rightarrow 0$. The relation to the V -states observed numerically in Wu et al. (J Comput Phys 53:1–42, 1984), Cerretelli and Williamson (J Fluid Mech 493:219–229, 2003) is discussed.

1. Introduction

In this paper, we study a certain class of the active scalar equations on the plane. Suppose we are given a function $D(z)$, $z = (x, y) \in \mathbb{R}^2$ that satisfies the following properties: D is radially symmetric, i.e., $D(z) = d(|z|)$, $d(r)$ is monotonically increasing and smooth on $(0, \infty)$. Consider the following transport equation

$$\dot{\theta} = \nabla \theta \cdot \left(\nabla^\perp A \theta + S \right), \quad \theta(0, x, y) = \theta_0(x, y), \quad (1)$$

where

$$A f = \int_{\mathbb{R}^2} D(z - \xi) f(\xi) d\xi, \quad z, \xi \in \mathbb{R}^2, \quad \nabla^\perp = (-\partial_y, \partial_x).$$

The symbol $S(t, z)$ will denote the strain, i.e., an exterior velocity that is assumed to be incompressible and sufficiently regular. By choosing different $d(r)$ and $S(z, t)$, we can cover some important cases. For example, taking $d(r) = -r^{-1}$ and $S(z, t) = 0$ corresponds to the so-called surface-quasigeostrophic equation (SQG), for which only the local in time solvability is known for sufficiently smooth θ_0 (see [4] for the recent development). If $d(r) = \log r$ and $S(z, t) = 0$, one recovers the equation for vorticity for

two-dimensional non-viscous Euler equation. In this situation, the existence of global solution $\theta(t, x, y)$ has been known for a long time [15].

In this paper, we mostly focus on the Euler case, however, we will be digressing to more general situations later in the text. Let us assume $\theta_0 \in L^\infty(\mathbb{R}^2) \cap L^1(\mathbb{R}^2)$. In that case, the existence of the global weak solution was established by Yudovich [17]. If $\theta_0 = \chi_{\Omega_0}$ with some domain $\Omega_0 : |\Omega_0| < \infty$, then one has the evolution of the ‘‘patch’’ as $\theta(t, z) = \chi_{\Omega(t)}$ and $\Omega(t)$ is homeomorphic to Ω_0 .

We consider the case when $\theta(0, z) = \chi_{\Omega_0}(z) + \chi_{-\Omega_0}(z)$ and $-\Omega_0 = \{-z, z \in \Omega_0\}$. Assuming $\Omega_0 \cap -\Omega_0 = \emptyset$, one has $\theta(t, z) = \chi_{\Omega(t)}(z) + \chi_{-\Omega(t)}(z)$, i.e., it represents evolution of the centrally symmetric pair of patches (the preservation of central symmetry is a simple feature of dynamics). We also take Ω_0 to be simply connected with smooth boundary, i.e., $\partial\Omega_0 \in C^\infty$. Under these assumptions, we have $\partial\Omega(t) \in C^\infty$ for all time [2,5].

Two problems arise naturally in the study of this model. The first one addresses the following question. Let $\vartheta(t) = \text{dist}(\Omega(t), 0)$. Is it possible for $\vartheta(t) \rightarrow 0, t \rightarrow \infty$? The Yudovich theory gives a lower bound: $\vartheta(t) > \exp(-\exp(Ct))$ and one can study its sharpness. Naturally, the convergence of $\vartheta(t)$ to zero implies the merging of the patches as the configuration is centrally symmetric. In the recent paper [8], the sharpness of the double exponential estimate was established (even up to a constant C) in the case when the equation was considered with the strain S , which was assumed to be incompressible, odd, and Lipschitz-regular. We have more evidence of the singularity formation in (1): in [10], the phenomenon of double exponential merging was proved for the Euler equation on the disc; however, the presence of the boundary was used in a substantial way. For the related SQG model, there is a numerical evidence that two patches merge in finite time [7].

The second problem, intimately related to the first one, is the existence of the quasi-stationary states, i.e., the configurations of two centrally symmetric patches that rotate with constant angular velocity around the origin without changing the shape (the so-called V -states). For the 2D Euler, there is numerical evidence (e.g., [3,16] and references there) for the existence of the parametric curve of these V -states: $V_\lambda \cup -V_\lambda$, i.e., $\Omega(t) = R_{\omega t} V_\lambda$, where R_θ denotes the rotation around the origin by the angle θ . Here, $\text{dist}(V_\lambda, -V_\lambda) = 2\lambda$ and $\lambda \in [0, \lambda_0)$. For $\lambda > 0$, the boundary $\Gamma_\lambda = \partial V_\lambda$ seems to be smooth but the two patches form a sharp corner of 90° and touch at the origin when $\lambda = 0$. Assuming existence of the V -states in the contact position and their regularity away from the origin, Overman [13] did a careful analysis around the zero. In particular, he explained why the 90° is the only possible nontrivial angle of the contact.

The paper consists of eight sections and an appendix. In the second section, the model with cut-off is introduced and the main result is stated. Section 3 gives some preliminaries and Sect. 4 explains the strategy of the proof. In the Sects. 5, 6, and 7, we prove some auxiliary statements that are used in Sect. 8 to complete the proofs of the main results. The appendix has four lemmas we use in the main text.

Notation used in the paper. The symbol $\dot{L}ip[0, T]$ will indicate the following space $\dot{L}ip[0, T] = \{f' \in L^\infty[0, T], f(0) = 0\}$ equipped with the norm

$$\|f\|_{\dot{L}ip[0, T]} = \sup_{x \in [0, T]} |f'(x)|.$$

We will use the following (non-standard) notation

$$\log^+ x = |\log x| + 1, \quad x > 0.$$

Let $\omega(x)$ be a smooth function such that $\omega(x) = 1$ on $|x| < 1/2$, $\omega(x) = 0$ on $|x| > 1$ and $0 \leq \omega(x) \leq 1$. For a parameter $a > 0$, we consider $\omega_a(x) = \omega(x/a)$ and $\omega_a^c(x) = 1 - \omega_a(x)$. Given two positive functions F_1 and F_2 , we write $F_1 \lesssim F_2$ if there is a constant C such that

$$F_1 < CF_2, \quad C > 0.$$

We write $F_1 \sim F_2$ if

$$F_1 \lesssim F_2 \lesssim F_1.$$

The expression “ $a \ll 1$ ” will be a short-hand for “there is a sufficiently small a_0 such that $0 < a < a_0$ ”. For the function $P(x)$, we write $\Delta_{x_1, x_2} P = P(x_1) - P(x_2)$.

2. The Model with Cut-Off and the Main Result

Consider $d(r) = \log r$ (2d Euler). If $\Omega_{sc}(0)$ is a simply connected domain with smooth boundary, the evolution of $\Gamma_{sc}(t) = \partial\Omega_{sc}(t)$ is governed by the following integro-differential equation (see, e.g., [1], formula (8.56); this is a corollary of $\nabla_z D(|z - \xi|) = -\nabla_\xi D(|z - \xi|)$ and the Green’s formula):

$$\dot{z}(t, \alpha) = C \int_0^{2\pi} z'(t, \beta) \log |z(t, \beta) - z(t, \alpha)| d\beta, \quad \alpha \in [0, 2\pi), \tag{2}$$

where C is an absolute constant and $z(\alpha, t)$ is anti-clockwise parameterization of the curve $\Gamma_{sc}(t)$. In particular, the right-hand side gives the velocity at any point on the boundary, $z(t, \alpha)$. If one has two simply connected domains $\Omega^{(1)}$ and $\Omega^{(2)}$ with vorticity equal to 1 inside each of them, then the velocity at any point $z_1 \in \Gamma^{(1)}$ is given by

$$C \left(\int_0^{2\pi} z'_1(t, \beta) \log |z_1(t, \beta) - z_1(t, \alpha)| d\beta + \int_0^{2\pi} z'_2(t, \beta) \log |z_2(t, \beta) - z_1(t, \alpha)| d\beta \right), \quad \alpha \in [0, 2\pi).$$

Assume now that two patches are merging at the origin. Then, we can introduce the local chart in $\{(x, y) : |x| < \delta, |y| < \delta\}$ and parameterize the corresponding contours by $(x, \mu_1(t, x))$ and $(x, \mu_2(t, x))$, see Fig. 1. Notice that the velocity at any point $(x, \mu_1(t, x))$ can now be written as

$$C \left(\int_{-\delta}^{\delta} (1, \mu'_1(t, \xi)) \log \left((x - \xi)^2 + (\mu_1(t, x) - \mu_1(t, \xi))^2 \right) d\xi - \int_{-\delta}^{\delta} (1, \mu'_2(t, \xi)) \log \left((x - \xi)^2 + (\mu_1(t, x) - \mu_2(t, \xi))^2 \right) d\xi \right) + R(x, \mu_1(t, x)),$$

where the negative sign in front of the second integral comes from the anti-clockwise parameterization for the contour $\Gamma^{(2)}$. Here R is the velocity induced by the $(\Gamma^{(1)} \cup \Gamma^{(2)}) \cap \{|z| > \delta\}$. Clearly, R is smooth inside $\{|x| < \delta/2, |y| < \delta/2\}$ and is equal to zero at the origin in the case when $\Omega^{(2)} = -\Omega^{(1)}$. Dropping this R as a term negligible around the origin, we end up with the following expression for the velocity

$$C \int_{-\delta}^{\delta} (A - B, \mu'_1(t, \xi)A - \mu'_2(t, \xi)B) d\xi,$$

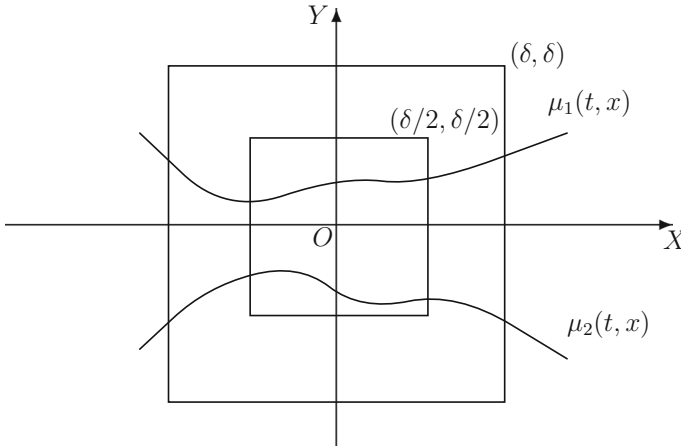


Fig. 1. Two patches near the origin

where

$$A = \log \left((x - \xi)^2 + (\mu_1(t, x) - \mu_1(t, \xi))^2 \right),$$

$$B = \log \left((x - \xi)^2 + (\mu_1(t, x) - \mu_2(t, \xi))^2 \right)$$

at every point $(x, \mu_1(t, x))$, $|x| < \delta$. Following, e.g., [14], we notice that the subtraction of any tangential vector from the velocity does not change the evolution of the contour. Thus, we subtract the vector-field

$$C \int_{-\delta}^{\delta} (A - B) d\xi \cdot (1, \mu'_1(t, x)),$$

which gives a modified velocity

$$u_{mod}(x, \mu_1(t, x)) = C \left(0, \int_{-\delta}^{\delta} (\mu'_1(t, \xi)A - \mu'_2(t, \xi)B - \mu'_1(t, x)A + \mu'_1(t, x)B) d\xi \right).$$

Notice that the first component of this vector is zero so we have an equation

$$\dot{\mu}_1(t, x) = C \int_{-\delta}^{\delta} (\mu'_1(t, \xi)A - \mu'_2(t, \xi)B - \mu'_1(t, x)A + \mu'_1(t, x)B) d\xi, \quad |x| < \delta/2 \tag{3}$$

for the evolution of $\mu_1(t, x)$. The similar formula can be obtained for $\mu_2(t, x)$. We will focus on the situation when the contours are centrally symmetric so $\mu_2(t, x) = -\mu_1(t, -x)$. That gives us the following equation for $\mu(t, x) = \mu_1(t, x)$:

$$\dot{\mu}(t, x) = C \int_{-\delta}^{\delta} (\mu'(t, x) - \mu'(t, \xi)) \log \left(\frac{(x + \xi)^2 + (\mu(t, x) + \mu(t, \xi))^2}{(x - \xi)^2 + (\mu(t, x) - \mu(t, \xi))^2} \right) d\xi. \tag{4}$$

We allow this equation to hold on all of $[-\delta, \delta]$ and call it *an equation with cut-off*. The rescaling of time makes it possible to adjust the value of C . Moreover, the Eq. (4) should be complemented by

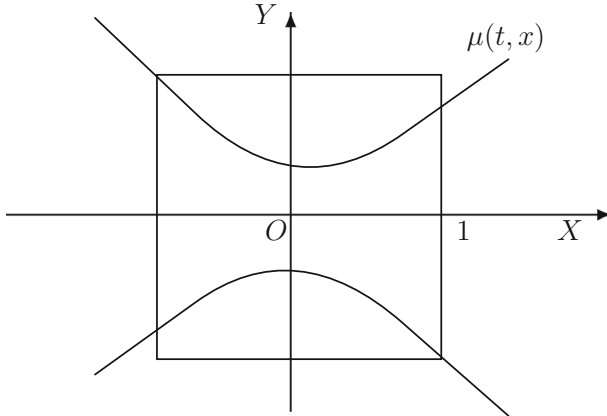


Fig. 2. Parametrization of the centrally symmetric pair near the origin

$$\mu(0, x) = \mu_0(x), \quad \mu(t, \delta) = c(t), \tag{5}$$

where $\mu_0(x)$ gives an initial position of the curve and $c(t)$ defines the *control* or the boundary value for this local transport equation.

In the case of the general kernel D in (1), the resulting equation can be reduced to the following form

$$\dot{\mu}(t, x) = C \int_{-\delta}^{\delta} (\mu'(t, x) - \mu'(t, \xi)) K(x, \xi) d\xi, \quad \mu(0, x) = \mu_0(x), \quad \mu(t, 1) = c(t), \tag{6}$$

where C is a floating constant (can be adjusted by time scaling),

$$K(x, \xi) = H((\mu(t, x) + \mu(t, \xi))^2 + (x + \xi)^2) - H((\mu(t, x) - \mu(t, \xi))^2 + (x - \xi)^2)$$

and $H(r) = d(\sqrt{r})$. If the function $d(r) = \log r$ or is homogeneous, the scaling in z allows one to assume that $\delta = 1$ which we will do from now on (see Fig. 2).

We think that problem with the cut-off might serve as a good model to study the merging of the central pair. Indeed, the active scalar equations are nonlocal but it is believed that the singularity of the convolution kernel at $r = 0$ is responsible for the strong instability (e.g., merging). That suggests a local version of the Eq. (2) might be interesting to study first. For this purpose, we take (6) as a model. The local in time solvability of (6) is not known and will be addressed elsewhere. However the “raison d’être” is different and can be formulated as the

Problem 1. Is there a smooth solution to

$$\dot{\mu}(t, x) = \int_{-1}^1 (\mu'(t, x) - \mu'(t, \xi)) K(x, \xi) d\xi, \quad x \in (-1, 1) \tag{7}$$

such that

$$\mu(t, x) \rightarrow y_0(x) = |x|, \quad t \rightarrow \infty.$$

If so, what are the estimates (lower and upper) on

$$\vartheta(t) = \min_{x \in [-1, 1]} |\mu(t, x) - y_0(x)|?$$

The function $y_0(x)$ plays a very special role. It is a stationary solution to (7) and it mimics locally the limiting case ($t = \infty$) considered in [8]. The advantage of the model (6) is that we know this singular stationary configuration exactly and the Problem 1 asks for the analysis of its dynamical stability. In particular, is it possible for $\vartheta(t)$ to converge to zero as double-exponential in the case when $d(r) = \log r$ (2d Euler)? In [8], this question was answered affirmatively assuming that a regular strain is allowed (see Appendix in [8]). In other words, an approximate solution to (7) was constructed and its self-similarity analysis was performed. The question remains though whether this strain can be dropped and this is the content of the Problem 1.

The Problem 1 seems hard. The important step in understanding it is to address the question of the stationary states for (7).

Problem 2. Find the family of even positive functions $y(x, \lambda) \in C^1[-1, 1]$ such that

$$\int_{-1}^1 (y'(x, \lambda) - y'(\xi, \lambda))K(x, \xi)d\xi = 0, \quad x \in [-1, 1] \tag{8}$$

and

$$y(0, \lambda) = \lambda, \quad \lambda \in (0, \lambda_0); \quad \|y(x, \lambda) - |x|\|_{C[-1,1]} \rightarrow 0, \quad \lambda \rightarrow 0.$$

Quite naturally, we will call these functions “the even V -states for the model with cut-off”. Since the original problem of the patch evolution is invariant with respect to rotations, we expect the existence of other families of V -states that are not necessarily even.

The main result of this paper is the following theorem which contains a solution to the Problem 2 for the case of 2d Euler.

Theorem 2.1. *There is a family of even positive functions $y(x, \lambda) \in C^1[-1, 1]$ such that*

$$\int_{-1}^1 (y'(x, \lambda) - y'(\xi, \lambda)) \log \left(\frac{(x + \xi)^2 + (y(x, \lambda) + y(\xi, \lambda))^2}{(x - \xi)^2 + (y(x, \lambda) - y(\xi, \lambda))^2} \right) d\xi = 0, \quad x \in [-1, 1] \tag{9}$$

and

$$y(0, \lambda) = \lambda, \quad \lambda \in (0, \lambda_0); \quad \lim_{\lambda \rightarrow 0} \|y(x, \lambda) - |x|\|_{C[-1,1]} = 0.$$

Remark. This result does not immediately imply any progress on Problem 1, however the developed technique might be useful.

Remark. The model with a cut-off we introduced is only a model, obviously. It is easier than the real Euler evolution on the plane because, e.g., it possesses the explicit singular steady state, $|x|$. However, in the case of 2d Euler (or SQG) equation on the torus $\mathbb{T}^2 = [-1, 1]^2$, the analog of $y_0(x)$ is the following configuration: $\theta_s(x, y) = \text{sign } x \cdot \text{sign } y$ which represents two patches that touch each other at the 90° angle. The method developed in this paper is likely to be directly applicable to the bifurcation analysis of this case which is NOT a model. We will address this issue elsewhere.

Remark. The bifurcation analysis of the stationary states is a classical subject in the mechanics of fluids (see, e.g., [6, 11] for the recent developments). We, however, focus on the technically hard case when the *singular* stationary state is considered.

3. Preliminaries

The main result of this paper is solution to Problem 2 in the case of 2d Euler equation with a cut-off. We start with some preliminary calculations for the general case as that will help us understand Problem 2 better.

Assume that $y(x, \lambda)$ solves the Problem 2. Since y is even in x , we have

$$y'(x, \lambda) \int_0^1 K_1(x, \xi) d\xi = \int_0^1 y'(\xi, \lambda) K_2(x, \xi) d\xi, \quad y(0, \lambda) = \lambda, \tag{10}$$

where

$$K_1(x, \xi) = K(x, \xi) + K(x, -\xi) = H((y(x) + y(\xi))^2 + (x + \xi)^2) - H((y(x) - y(\xi))^2 + (x - \xi)^2) + H((y(x) + y(\xi))^2 + (x - \xi)^2) - H((y(x) - y(\xi))^2 + (x + \xi)^2)$$

and

$$K_2(x, \xi) = K(x, \xi) - K(x, -\xi) = H((y(x) + y(\xi))^2 + (x + \xi)^2) - H((y(x) - y(\xi))^2 + (x - \xi)^2) + (x - \xi)^2 - H((y(x) + y(\xi))^2 + (x - \xi)^2) + H((y(x) - y(\xi))^2 + (x + \xi)^2).$$

We suppress the dependence of y on λ and just write $y(x)$ here.

3.1. The explicit solution for the model case. Let us go back to the Eq. (1). Instead of taking the singular kernels in the convolution, one can consider the smooth bump $D(z)$. The ‘‘typical’’ behavior around the origin then would be, e.g.,

$$D(z) = C + |z|^2 + o(|z|^2), \quad |z| \rightarrow 0.$$

Keeping only the quadratic part, we get

$$K(x, \xi) = 4(y(x)y(\xi) + x\xi), \quad K_1(x, \xi) = 8y(x)y(\xi), \quad K_2(x, \xi) = 8x\xi.$$

The Eq. (7) takes the following form

$$y'(x)y(x) \int_0^1 y(\xi) d\xi = x \int_0^1 \xi y'(\xi) d\xi$$

which easily integrates to

$$y(x) = \sqrt{\lambda^2 + \frac{B}{A}x^2},$$

where

$$A = \int_0^1 y(x) dx, \quad B = \int_0^1 xy'(x) dx.$$

We have the following compatibility equations

$$\left\{ \begin{array}{l} B = \sqrt{\lambda^2 + \frac{B}{A}} - A \\ A = \int_0^1 \sqrt{\lambda^2 + \frac{B}{A}x^2} dx \end{array} \right\} \left\{ \begin{array}{l} B = \sqrt{\lambda^2 + \frac{B}{A}} - A \\ \sqrt{AB} = \lambda^2 \int_0^{\lambda^{-1}\sqrt{BA^{-1}}} \sqrt{1 + \xi^2} d\xi \end{array} \right. .$$

Introduce

$$B/A = u, \quad AB = v.$$

Then

$$v = \frac{u(\lambda^2 + u)}{(u + 1)^2}, \quad \sqrt{v} = \lambda^2 \int_0^{\lambda^{-1}\sqrt{u}} \sqrt{1 + \xi^2} d\xi.$$

We assume that $\lambda \in (0, \lambda_0)$, $\lambda_0 \ll 1$ and $|u - 1| \ll 1$ and so $|v - 1/4| \ll 1$. Therefore, if

$$u = 1 + \alpha, \quad v = 1/4 + \beta, \quad \alpha, \beta \ll 1$$

then

$$\beta = \alpha/4 + \lambda^2/4 + O(\alpha^2 + \lambda^2\alpha)$$

and

$$\alpha = 2\beta - \lambda^2 \log \frac{1}{\lambda} + O(\beta^2 + \lambda^2\alpha).$$

Thus,

$$\beta = -0.5\lambda^2 \log \frac{1}{\lambda} + \frac{\lambda^2}{2} + O(\lambda^4 \log^2 \lambda), \quad \alpha = -2\lambda^2 \log \frac{1}{\lambda} + \lambda^2 + O(\lambda^4 \log^2 \lambda)$$

This calculation shows that V_λ exists and the asymptotics in $\lambda \rightarrow 0$ can be easily established. Since

$$y(x) = \sqrt{\lambda^2 + (1 + \alpha)x^2}, \quad \alpha < 0$$

the curve will intersect the line $y = x$ at the point

$$x_\lambda^* = \frac{\lambda}{|\alpha|^{1/2}} = \left(2 \log \frac{1}{\lambda}\right)^{-1/2} (1 + o(1)).$$

Now, let us address the question of self-similarity. Rescale

$$\mu(\widehat{x}) = \lambda^{-1}y(\widehat{x}\lambda) = \sqrt{1 + (1 + \alpha)\widehat{x}^2}, \quad |\widehat{x}| < \lambda^{-1}.$$

This shows that

$$\sup_{|\widehat{x}| < \lambda^{-1}} |\mu(\widehat{x}) - \sqrt{1 + \widehat{x}^2}| \rightarrow 0$$

and so the self-similar behavior is global.

The model case we just considered is the situation in which the interaction is substantially long-range and the self-similarity of the stationary state is global. The curve that we have in the limit is hyperbola. That seems like a common feature of many long-range models and 2d Euler in particular as will be seen from the subsequent analysis. However, for 2d Euler this self-similarity will be proved only over $|x| < C\lambda$ with arbitrary fixed C . Notice also that the analogous calculation is possible if the smooth strains are imposed, e.g., a rotation.

3.2. *Properties of the kernels K_1 and K_2 .* Below, we will write $K_{1(2)}(x, \xi, y)$ when we want to emphasize the dependence of the kernel on the function y .

Lemma 3.1. *The following is true*

$K_1(x, \xi, y) = 4y(x)y(\xi)(H'(\eta_1) + H'(\eta_2)), K_2(x, \xi, y) = 4x\xi(H'(\alpha_1) + H'(\alpha_2)),$
 where

$$\eta_1 > (x + \xi)^2, \eta_2 > (x - \xi)^2$$

and

$$\alpha_{1(2)} > (x - \xi)^2.$$

Proof. Apply the mean value theorem to the first and second terms in the expression. This gives

$$K_1 = \left(H((y(x) + y(\xi))^2 + (x + \xi)^2) - H((y(x) - y(\xi))^2 + (x + \xi)^2) \right) + \left(H((y(x) + y(\xi))^2 + (x - \xi)^2) - H((y(x) - y(\xi))^2 + (x - \xi)^2) \right)$$

and

$$K_2 = \left(H((y(x) + y(\xi))^2 + (x + \xi)^2) - H((y(x) + y(\xi))^2 + (x - \xi)^2) \right) + \left(H((y(x) - y(\xi))^2 + (x + \xi)^2) - H((y(x) - y(\xi))^2 + (x - \xi)^2) \right).$$

□

If $H = \log x$, we have the following representations

$$K_1(x, \xi, y) = \log \left(\frac{(x + \xi)^2 + (y(x) + y(\xi))^2}{(x - \xi)^2 + (y(x) - y(\xi))^2} \cdot \frac{(x - \xi)^2 + (y(x) + y(\xi))^2}{(x + \xi)^2 + (y(x) - y(\xi))^2} \right)$$

and

$$K_2(x, \xi, y) = \log \left(\frac{(x + \xi)^2 + (y(x) + y(\xi))^2}{(x - \xi)^2 + (y(x) + y(\xi))^2} \cdot \frac{(x + \xi)^2 + (y(x) - y(\xi))^2}{(x - \xi)^2 + (y(x) - y(\xi))^2} \right).$$

Then, assuming that $y(x) \geq 0$,

$$\frac{(x - \xi)^2 + (y(x) + y(\xi))^2}{(x - \xi)^2 + (y(x) - y(\xi))^2} = 1 + \frac{4y(x)y(\xi)}{(x - \xi)^2 + (y(x) - y(\xi))^2} \geq 1$$

and

$$\frac{(x + \xi)^2 + (y(x) + y(\xi))^2}{(x + \xi)^2 + (y(x) - y(\xi))^2} = 1 + \frac{4y(x)y(\xi)}{(x + \xi)^2 + (y(x) - y(\xi))^2} \geq 1.$$

Therefore, we have

$$\log \left(1 + \frac{4y(x)y(\xi)}{(x - \xi)^2 + (y(x) - y(\xi))^2} \right) \leq K_1 \lesssim \log \left(1 + \frac{4y(x)y(\xi)}{(x - \xi)^2} \right),$$

provided that $y \geq 0$. Similarly, for K_2 ,

$$\begin{aligned} & \log \left(1 + \frac{4x\xi}{(x - \xi)^2 + (y(x) - y(\xi))^2} \right) \\ & \leq K_2 = \log \left(1 + \frac{4x\xi}{(x - \xi)^2 + (y(x) + y(\xi))^2} \right) + \log \left(1 + \frac{4x\xi}{(x - \xi)^2 + (y(x) - y(\xi))^2} \right) \\ & \lesssim \log \left(1 + \frac{4x\xi}{(x - \xi)^2} \right) \end{aligned}$$

and this holds for all y .

The following lemma is trivial.

Lemma 3.2. *Let $0 \leq a \leq b \leq C$. Then, $b \sim a + b$ and*

$$\frac{1}{b - a} \int_a^b \frac{d\eta}{\eta} \sim b^{-1} \sim (a + b)^{-1}, \quad a > b/2$$

and

$$\frac{1}{a + b} \lesssim \frac{1}{b - a} \int_a^b \frac{d\eta}{\eta} \lesssim \frac{\log^+ a}{a + b}, \quad a < b/2.$$

Suppose that $y \in [0, C]$. Then, applying this lemma to K_1 with

$$a = (y(x) - y(\xi))^2 + (x + \xi)^2, \quad b = (y(x) + y(\xi))^2 + (x + \xi)^2$$

and then with

$$a = (y(x) - y(\xi))^2 + (x - \xi)^2, \quad b = (y(x) + y(\xi))^2 + (x - \xi)^2$$

gives

$$\frac{y(x)y(\xi)}{y^2(x) + y^2(\xi) + (x - \xi)^2} \lesssim K_1 \lesssim y(x)y(\xi) \frac{\log^+((x - \xi)^2 + (y(x) - y(\xi))^2)}{y^2(x) + y^2(\xi) + (x - \xi)^2} \tag{11}$$

For K_2 , the same reasoning yields

$$\frac{x\xi}{x^2 + \xi^2 + (y(x) - y(\xi))^2} \lesssim K_2 \lesssim x\xi \frac{\log^+((x - \xi)^2 + (y(x) - y(\xi))^2)}{y^2(x) + y^2(\xi) + (x - \xi)^2}.$$

4. The Implicit Function Theorem, the 2d Euler Case

In this section, we will apply the scheme of the implicit function theorem to the 2d Euler with cut-off which corresponds to $H(x) = \log x$. However, we first notice that the problem allows the following scaling.

Lemma 4.1. *If $y(x)$ solves*

$$\begin{aligned} & \int_{-1}^1 y'(x) \log \left(\frac{(x + \xi)^2 + (y(x) + y(\xi))^2}{(x - \xi)^2 + (y(x) - y(\xi))^2} \right) d\xi \\ & = \int_{-1}^1 y'(\xi) \log \left(\frac{(x + \xi)^2 + (y(x) + y(\xi))^2}{(x - \xi)^2 + (y(x) - y(\xi))^2} \right) d\xi \end{aligned}$$

then $y_\alpha(x) = \alpha y(x/\alpha)$ solves

$$\begin{aligned} & \int_{-\alpha}^{\alpha} y'_\alpha(x) \log \left(\frac{(x + \xi)^2 + (y_\alpha(x) + y_\alpha(\xi))^2}{(x - \xi)^2 + (y_\alpha(x) - y_\alpha(\xi))^2} \right) d\xi \\ &= \int_{-\alpha}^{\alpha} y'_\alpha(\xi) \log \left(\frac{(x + \xi)^2 + (y_\alpha(x) + y_\alpha(\xi))^2}{(x - \xi)^2 + (y_\alpha(x) - y_\alpha(\xi))^2} \right) d\xi \end{aligned}$$

for every $\alpha > 0$.

Proof. The proof is an immediate calculation. \square

Consider $y_\lambda(x)$ and take

$$\widehat{y}(\widehat{x}, \lambda) = \lambda^{-1} y(\widehat{x}\lambda, \lambda), \quad |\widehat{x}| < \lambda^{-1}.$$

We will perform this scaling many times in the paper. It allows to reduce the problem to the one on the larger interval $|\widehat{x}| < \lambda^{-1}$ with the normalization $\widehat{y}(0, \lambda) = 1$.

Remark 2. The perturbative analysis done below will be carried out around the hyperbola $\widehat{y}(\widehat{x}) = \sqrt{\widehat{x}^2 + 1}$, not $|\widehat{x}|$. The explanation to that is the following. The model case suggests that $\{\widehat{y}(\widehat{x}, \lambda)\}$ might have some limiting behavior as $\lambda \rightarrow 0$. If so, can one guess the asymptotical curve? To this end, let us make very natural assumptions that

$$\widehat{y}(\widehat{x}, \lambda) \rightarrow f(\widehat{x}), \quad \widehat{y}'(\widehat{x}, \lambda) \rightarrow f'(\widehat{x})$$

on every interval $\widehat{x} \in [-C, C]$ and that

$$\widehat{y}(\widehat{x}, \lambda) = \widehat{x}(1 + o(1)), \quad \widehat{y}'(\widehat{x}, \lambda) = 1 + o(1), \quad |\widehat{x}| \gg 1$$

uniformly in $\lambda \in (0, \lambda_0]$. For $|\widehat{x}| < C$,

$$\begin{aligned} & (f'(\widehat{x}) + o(1)) \int_0^{1/\lambda} \left[\log \left(1 + \frac{4\widehat{y}(\widehat{x}, \lambda)\widehat{y}(\widehat{\xi}, \lambda)}{(\widehat{x} - \widehat{\xi})^2 + (\widehat{y}(\widehat{x}, \lambda) - \widehat{y}(\widehat{\xi}, \lambda))^2} \right) \right. \\ & \quad \left. + \log \left(1 + \frac{4\widehat{y}(\widehat{x}, \lambda)\widehat{y}(\widehat{\xi}, \lambda)}{(\widehat{x} + \widehat{\xi})^2 + (\widehat{y}(\widehat{x}, \lambda) - \widehat{y}(\widehat{\xi}, \lambda))^2} \right) \right] d\widehat{\xi} \\ &= \int_0^{1/\lambda} (1 + o(1)) \left[\log \left(1 + \frac{4\widehat{x}\widehat{\xi}}{(\widehat{x} - \widehat{\xi})^2 + (\widehat{y}(\widehat{x}, \lambda) - \widehat{y}(\widehat{\xi}, \lambda))^2} \right) \right. \\ & \quad \left. + \log \left(1 + \frac{4\widehat{x}\widehat{\xi}}{(\widehat{x} + \widehat{\xi})^2 + (\widehat{y}(\widehat{x}, \lambda) - \widehat{y}(\widehat{\xi}, \lambda))^2} \right) \right] d\widehat{\xi} \end{aligned}$$

For the l.h.s., the asymptotics of the integrand as $\widehat{\xi} \rightarrow \infty$ is

$$\frac{4\widehat{y}(\widehat{x}, \lambda)}{\widehat{\xi}} + o(\widehat{\xi}^{-1})$$

and for the r.h.s., it is

$$\frac{4\widehat{x}}{\widehat{\xi}} + o(\widehat{\xi}^{-1}).$$

Here we work under assumption that $|\widehat{x}| < C$. Taking $\lambda \rightarrow 0$, we get

$$(f' f - \widehat{x}) \log(1/\lambda) + o(\log(1/\lambda)) = 0.$$

This leads to $f' f - \widehat{x} = 0$ and (since $f(0) = 1$)

$$f(\widehat{x}) = (\widehat{x}^2 + 1)^{1/2} \tag{12}$$

This formula was obtained under strong assumptions so does not imply the self-similarity per se. However, one can take

$$\widetilde{y}(x, \lambda) = (x^2 + \lambda^2)^{1/2}$$

as an approximate solution. Plugging it into the equation, one can represent the resulting correction as the strain. Similarly to [8], one can show that this strain satisfies the uniform bound

$$\sup_{|z| < 1, \lambda \in (0, 1)} \frac{|S(z, \lambda)|}{|z|} < C.$$

The novelty of the current paper is that we construct the *exact* solution and thus make $S(z, \lambda) = 0$. It will also be proved that the exact solutions converge to hyperbola in the scaling limit but only locally, over $x \in I_\lambda$, where $|I_\lambda| \rightarrow 0, \lambda \rightarrow 0$.

In the lemma below, we show that all possible solutions $y(x, \lambda)$ have the following common feature.

Lemma 4.2. *If $y(x)$ solves (10), then there is $x^* \in (0, 1)$ at which $y(x^*) = x^*$. That is, the graph of $y(x)$ intersects the line $y = x$.*

Proof. Suppose instead that $y(x) > x$ for all $x \in (0, 1)$. Then,

$$\frac{4y(x)y(\xi)}{(x - \xi)^2 + (y(x) - y(\xi))^2} > \frac{4x\xi}{(x - \xi)^2 + (y(x) - y(\xi))^2}$$

and

$$\frac{4y(x)y(\xi)}{(x + \xi)^2 + (y(x) - y(\xi))^2} > \frac{4x\xi}{(x - \xi)^2 + (y(x) + y(\xi))^2}.$$

Therefore, $K_1(x, \xi) > K_2(x, \xi) > 0$. Now, assume that

$$\max_{x \in [0, 1]} y'(x) = y'(x_1)$$

Then,

$$\int_0^1 y'(\xi) K_1(x_1, \xi) d\xi \leq y'(x_1) \int_0^1 K_1(x_1, \xi) d\xi = \int_0^1 y'(\xi) K_2(x_1, \xi) d\xi$$

and this inequality is strict unless $y'(x) = \text{const}$. This is impossible by, e.g., the smoothness assumption. \square

Now that we established what properties the solution $y(x, \lambda)$ needs to possess, we are ready to prove its existence.

Consider small $\delta > 0$ and the sets

$$\Omega = \{f : \|f(x) - x\|_{Lip[0,1]} \leq \delta\}, \quad I = \{\lambda : \lambda \in (0, \lambda_0], \quad \lambda_0 \ll 1\}.$$

We will look for $y = \sqrt{\lambda^2 + f^2(x)}$, where $(f, \lambda) \in \Omega \times I$. Notice that $f(x) = \int_0^x f'(t)dt$ and $\|f' - 1\|_{L^\infty[0,1]} \ll 1$. Therefore,

$$f(x) = x(1 + O(\delta)).$$

In particular, $f(x) > 0$ for $x > 0$.

Consider the functional (we specify the dependence of $K_{1(2)}$ on y here)

$$F(f, \lambda) = \frac{ff' \int_0^1 K_1(x, \tau, y)d\tau - \sqrt{\lambda^2 + f^2(x)} \int_0^1 y'(\tau)K_2(x, \tau, y)d\tau}{x\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)}$$

where $y = \sqrt{\lambda^2 + f^2(x)}$

which acts from $\Omega \times I$ to $L^\infty[0, 1]$. Moreover, $F(x, 0) = 0$.

The Eq. (10) can be rewritten as

$$F(f, \lambda) = 0.$$

We will solve it in the following way (this is essentially the implicit function theorem proof [9] but we prefer to give the argument for the sake of completeness). Write

$$F(f, \lambda) = F(x, \lambda) + (D_f F(x, \lambda))\psi + Q(\psi)$$

where $\psi = f - x$ and this representation defines an operator Q . That can be rewritten as

$$\psi = -\left(D_f F(x, \lambda)\right)^{-1} Q(\psi) + \psi_0(\lambda), \quad \psi_0 = -\left(D_f F(x, \lambda)\right)^{-1} F(x, \lambda). \quad (13)$$

Next, we will show that this equation can be solved by contraction mapping principle in $\mathcal{B}_\delta = \{\|\psi\|_{Lip[0,1]} \leq \delta\}$, $\delta \ll 1$. To this end, we only need to prove:

(a) Linear part:

$$\|(D_f F(x, \lambda))^{-1}\|_{L^\infty[0,1], Lip[0,1]} < \widehat{C} \quad (14)$$

if $\lambda \in (0, \lambda_1)$ with $\lambda_1 \ll 1$.

(b) Frechet differentiability:

$$\|Q(\psi)\|_{L^\infty[0,1]} = o(1)\|\psi\|_{Lip[0,1]} \quad (15)$$

and

$$\|Q(\psi_2) - Q(\psi_1)\|_{L^\infty[0,1]} = o(1)\|\psi_2 - \psi_1\|_{Lip[0,1]} \quad (16)$$

with $o(1) \rightarrow 0$ as $\delta \rightarrow 0$ uniformly in $\lambda \in (0, 1)$ and $\psi, \psi_{1(2)} \in \mathcal{B}_\delta$.

(c) Small initial data:

$$\|\psi_0(\lambda)\|_{Lip[0,1]} < \delta/2 \tag{17}$$

where $\lambda \in [0, \lambda_0], \lambda_0 < \lambda_1$.

We will first make λ_1 so small that (a) holds. Then, we choose δ small enough to have $o(1)$ in (b) at most $(10\widehat{C})^{-1}$ uniformly in $\lambda \in (0, 1)$. Finally, we take λ_0 so small that (c) holds. This will ensure existence and uniqueness of solution in the complete metric space \mathcal{B}_δ . Then, it will be easy to bootstrap its regularity from $Lip[-1, 1]$ to $C^1[-1, 1]$. The continuous dependence on λ and

$$\|y(x, \lambda) - x\|_{C[0,1]} \rightarrow 0, \quad \lambda \rightarrow 0$$

will follow from the proof.

5. The Analysis of Gateaux Derivative for $H(x)=\log x$

Taking $f_t = f + tu, u \in \dot{Lip}[0, 1]$, plugging it into F , and computing the derivative in t at $t = 0$ with positive x fixed, results in

$$(D_f F(f, \lambda))u = \frac{1}{x\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} (I_1 + \dots + I_6). \tag{18}$$

We have

$$\begin{aligned} I_1 &= \left(f' \int_0^1 K_1(x, \tau, y) d\tau \right) u, \quad y = \sqrt{\lambda^2 + f^2} \\ I_2 &= \left(f \int_0^1 K_1(x, \tau, y) d\tau \right) u' \\ I_3 &= ff' \int_0^1 \delta K_1(x, \tau, y) d\tau \end{aligned}$$

where

$$\begin{aligned} \delta K_1 &= \frac{2(y(x) + y(\xi))(\delta y(x) + \delta y(\xi))}{(x + \xi)^2 + (y(x) + y(\xi))^2} - \frac{2(y(x) - y(\xi))(\delta y(x) - \delta y(\xi))}{(x - \xi)^2 + (y(x) - y(\xi))^2} \\ &+ \frac{2(y(x) + y(\xi))(\delta y(x) + \delta y(\xi))}{(x - \xi)^2 + (y(x) + y(\xi))^2} - \frac{2(y(x) - y(\xi))(\delta y(x) - \delta y(\xi))}{(x + \xi)^2 + (y(x) - y(\xi))^2} \end{aligned}$$

and

$$\begin{aligned} \delta y &= \frac{fu}{\sqrt{\lambda^2 + f^2}} \\ I_4 &= - \left(\frac{f}{\sqrt{\lambda^2 + f^2}} \int_0^1 y' K_2(x, \tau, y) d\tau \right) u \\ I_5 &= -\sqrt{\lambda^2 + f^2} \int_0^1 \delta y' K_2(x, \tau, y) d\tau \end{aligned}$$

where

$$\delta y' = \frac{f'}{\sqrt{\lambda^2 + f^2}}u + \frac{f}{\sqrt{\lambda^2 + f^2}}u' - \frac{f^2 f' u}{(\lambda^2 + f^2)^{3/2}}$$

$$I_6 = -\sqrt{\lambda^2 + f^2} \int_0^1 y'(\tau) \delta K_2(x, \tau, y) d\tau$$

where

$$\delta K_2 = \frac{2(y(x) + y(\xi))(\delta y(x) + \delta y(\xi))}{(x + \xi)^2 + (y(x) + y(\xi))^2} - \frac{2(y(x) - y(\xi))(\delta y(x) - \delta y(\xi))}{(x - \xi)^2 + (y(x) - y(\xi))^2}$$

$$- \frac{2(y(x) + y(\xi))(\delta y(x) + \delta y(\xi))}{(x - \xi)^2 + (y(x) + y(\xi))^2} + \frac{2(y(x) - y(\xi))(\delta y(x) - \delta y(\xi))}{(x + \xi)^2 + (y(x) - y(\xi))^2}.$$

In the formulas above, we have

$$K_1(x, \xi, y) = \log \left(\frac{(x + \xi)^2 + (y(x) + y(\xi))^2}{(x - \xi)^2 + (y(x) - y(\xi))^2} \cdot \frac{(x - \xi)^2 + (y(x) + y(\xi))^2}{(x + \xi)^2 + (y(x) - y(\xi))^2} \right)$$

and

$$K_2(x, \xi, y) = \log \left(\frac{(x + \xi)^2 + (y(x) + y(\xi))^2}{(x - \xi)^2 + (y(x) + y(\xi))^2} \cdot \frac{(x + \xi)^2 + (y(x) - y(\xi))^2}{(x - \xi)^2 + (y(x) - y(\xi))^2} \right).$$

5.1. *The derivative at $f(x) = x$.* Define $L_\lambda = (D_f F)(x, \lambda)$. If $f = x$ in the previous section, then

$$L_\lambda u = \frac{1}{x\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} (\widehat{I}_{1,\lambda} + \dots + \widehat{I}_{6,\lambda}).$$

We again have

$$\widehat{I}_{1,\lambda} = \left(\int_0^1 K_1(x, \tau, y_\lambda) d\tau \right) u$$

with

$$y_\lambda(x) = \sqrt{\lambda^2 + x^2}$$

$$\widehat{I}_{2,\lambda} = x \left(\int_0^1 K_1(x, \tau, y_\lambda) d\tau \right) u'$$

$$\widehat{I}_{3,\lambda} = x \int_0^1 \delta K_1(x, \tau, y_\lambda) d\tau$$

where

$$\delta K_1 = \frac{2(y_\lambda(x) + y_\lambda(\xi))(\delta y_\lambda(x) + \delta y_\lambda(\xi))}{(x + \xi)^2 + (y_\lambda(x) + y_\lambda(\xi))^2} - \frac{2(y_\lambda(x) - y_\lambda(\xi))(\delta y_\lambda(x) - \delta y_\lambda(\xi))}{(x - \xi)^2 + (y_\lambda(x) - y_\lambda(\xi))^2}$$

$$+ \frac{2(y_\lambda(x) + y_\lambda(\xi))(\delta y_\lambda(x) + \delta y_\lambda(\xi))}{(x - \xi)^2 + (y_\lambda(x) + y_\lambda(\xi))^2} - \frac{2(y_\lambda(x) - y_\lambda(\xi))(\delta y_\lambda(x) - \delta y_\lambda(\xi))}{(x + \xi)^2 + (y_\lambda(x) - y_\lambda(\xi))^2}$$

and

$$\begin{aligned} \delta y_\lambda &= \frac{x}{\sqrt{\lambda^2 + x^2}} u \\ \widehat{I}_{4,\lambda} &= - \left(\frac{x}{\sqrt{\lambda^2 + x^2}} \int_0^1 y'_\lambda K_2(x, \tau, y_\lambda) d\tau \right) u \\ \widehat{I}_{5,\lambda} &= -\sqrt{\lambda^2 + x^2} \int_0^1 \delta y'_\lambda K_2(x, \tau, y_\lambda) d\tau \end{aligned}$$

where

$$\begin{aligned} \delta y'_\lambda &= \frac{1}{\sqrt{\lambda^2 + x^2}} u + \frac{x}{\sqrt{\lambda^2 + x^2}} u' - \frac{x^2}{(\lambda^2 + x^2)^{3/2}} u \\ \widehat{I}_{6,\lambda} &= -\sqrt{\lambda^2 + x^2} \int_0^1 y'_\lambda(\tau) \delta K_2(x, \tau, y_\lambda) d\tau \end{aligned}$$

and

$$\begin{aligned} \delta K_2 &= \frac{2(y_\lambda(x) + y_\lambda(\xi))(\delta y_\lambda(x) + \delta y_\lambda(\xi))}{(x + \xi)^2 + (y_\lambda(x) + y_\lambda(\xi))^2} - \frac{2(y_\lambda(x) - y_\lambda(\xi))(\delta y_\lambda(x) - \delta y_\lambda(\xi))}{(x - \xi)^2 + (y_\lambda(x) - y_\lambda(\xi))^2} \\ &\quad - \frac{2(y_\lambda(x) + y_\lambda(\xi))(\delta y_\lambda(x) + \delta y_\lambda(\xi))}{(x - \xi)^2 + (y_\lambda(x) + y_\lambda(\xi))^2} + \frac{2(y_\lambda(x) - y_\lambda(\xi))(\delta y_\lambda(x) - \delta y_\lambda(\xi))}{(x + \xi)^2 + (y_\lambda(x) - y_\lambda(\xi))^2}. \end{aligned}$$

5.2. *The operator L_λ .* For L_λ , we have the following formula

$$L_\lambda = A_1 u' + A_2 u + \int_0^1 D_1(x, \xi, \lambda) u(\xi) d\xi + \int_0^1 D_2(x, \xi, \lambda) u'(\xi) d\xi.$$

The equation

$$L_\lambda u = g$$

can be rewritten as

$$A_1(x, \lambda) u' + A_2(x, \lambda) u + \int_0^1 M(x, \xi, \lambda) u'(\xi) d\xi = g \tag{19}$$

if one assumes $u(0) = 0$ and

$$M(x, \xi, \lambda) = D_2(x, \xi, \lambda) + \int_\xi^1 D_1(x, \tau, \lambda) d\tau.$$

In the calculation above, we used

$$\lim_{x \rightarrow 0} \left(u(x) \int_0^1 D_1(x, \tau, \lambda) d\tau \right) = 0.$$

This equality follows from the estimate $|u(x)| \lesssim x$ and from the analysis of

$$\int_0^1 D_1(x, \tau, \lambda) d\tau$$

when $x \rightarrow 0$ (see (46) below).

Let us introduce the integral operator \mathcal{M}_λ with the kernel $M(x, \tau, \lambda)$, e.g.,

$$\mathcal{M}_\lambda f = \int_0^1 M(x, \tau, \lambda) f(\tau) d\tau$$

For the coefficients, we have

$$A_1 = \frac{\int_0^1 K_1(x, \tau, y_\lambda) d\tau}{\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)}.$$

The expression for A_2 is more complicated,

$$A_2 = \frac{1}{x\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} \left(\int_0^1 K_1(x, \tau, y_\lambda) d\tau - \frac{x}{\sqrt{\lambda^2 + x^2}} \int_0^1 y'_\lambda(\tau) K_2(x, \tau, y_\lambda) d\tau + B_2 \right), \tag{20}$$

where

$$B_2 = \frac{2x}{\sqrt{x^2 + \lambda^2}} \int_0^1 \left(x - \frac{\xi\sqrt{\lambda^2 + x^2}}{\sqrt{\lambda^2 + \xi^2}} \right) \left(\frac{y_\lambda(x) + y_\lambda(\xi)}{(x + \xi)^2 + (y_\lambda(x) + y_\lambda(\xi))^2} - \frac{y_\lambda(x) - y_\lambda(\xi)}{(x - \xi)^2 + (y_\lambda(x) - y_\lambda(\xi))^2} \right) d\xi + \frac{2x}{\sqrt{x^2 + \lambda^2}} \int_0^1 \left(x + \frac{\xi\sqrt{\lambda^2 + x^2}}{\sqrt{\lambda^2 + \xi^2}} \right) \left(\frac{y_\lambda(x) + y_\lambda(\xi)}{(x - \xi)^2 + (y_\lambda(x) + y_\lambda(\xi))^2} - \frac{y_\lambda(x) - y_\lambda(\xi)}{(x + \xi)^2 + (y_\lambda(x) - y_\lambda(\xi))^2} \right) d\xi.$$

For $D_{1(2)}$, one has

$$D_2(x, \xi, \lambda) = -\frac{1}{x \log^+(x^2 + \lambda^2)} K_2(x, \xi, y_\lambda) \frac{\xi}{\sqrt{\lambda^2 + \xi^2}}$$

and

$$D_1(x, \xi, \lambda) = \frac{1}{x\sqrt{\lambda^2 + x^2} \log^+(x^2 + \lambda^2)} \left[\frac{2x\xi}{\sqrt{\lambda^2 + \xi^2}} \left(\frac{y_\lambda(x) + y_\lambda(\xi)}{(x + \xi)^2 + (y_\lambda(x) + y_\lambda(\xi))^2} + \frac{y_\lambda(x) - y_\lambda(\xi)}{(x - \xi)^2 + (y_\lambda(x) - y_\lambda(\xi))^2} \right) + \frac{y_\lambda(x) - y_\lambda(\xi)}{(x + \xi)^2 + (y_\lambda(x) - y_\lambda(\xi))^2} \right] - \frac{2\xi^2\sqrt{\lambda^2 + x^2}}{\xi^2 + \lambda^2} \left(\frac{y_\lambda(x) + y_\lambda(\xi)}{(x + \xi)^2 + (y_\lambda(x) + y_\lambda(\xi))^2} + \frac{y_\lambda(x) - y_\lambda(\xi)}{(x - \xi)^2 + (y_\lambda(x) - y_\lambda(\xi))^2} \right) - \frac{y_\lambda(x) + y_\lambda(\xi)}{(x - \xi)^2 + (y_\lambda(x) + y_\lambda(\xi))^2} - \frac{y_\lambda(x) - y_\lambda(\xi)}{(x + \xi)^2 + (y_\lambda(x) - y_\lambda(\xi))^2} \left. - \frac{\lambda^2\sqrt{\lambda^2 + x^2}}{(\lambda^2 + \xi^2)^{3/2}} K_2(x, \xi, y_\lambda) \right].$$

In this section, we will obtain estimates/asymptotics of all four terms in the case when $\lambda \rightarrow 0$. It will be trivial to do that away from 0: e.g., for every $\delta > 0$ both $A_{1(2)}(\lambda) \rightarrow A_{1(2)}(0)$ uniformly over $x \in [\delta, 1]$. The behavior around 0 is delicate and will require more careful treatment.

We start with the following calculation that will simplify the expressions above.

We write

$$\sqrt{\widehat{x}^2 + 1} - \sqrt{\widehat{\xi}^2 + 1} = (\widehat{x} - \widehat{\xi})r_1(x, \xi) \tag{21}$$

where

$$r_1 = \frac{\widehat{x} + \widehat{\xi}}{\sqrt{\widehat{x}^2 + 1} + \sqrt{\widehat{\xi}^2 + 1}} = 1 + O\left(\frac{1}{\widehat{x}\widehat{\xi}}\right), \text{ if } \widehat{x}, \widehat{\xi} \gg 1. \tag{22}$$

Similarly,

$$\sqrt{\widehat{x}^2 + 1} + \sqrt{\widehat{\xi}^2 + 1} = (\widehat{x} + \widehat{\xi})r_1^{-1}, \quad r_1^{-1} = \frac{\sqrt{\widehat{x}^2 + 1} + \sqrt{\widehat{\xi}^2 + 1}}{\widehat{x} + \widehat{\xi}}. \tag{23}$$

Thus, we have for K_2

$$\begin{aligned} & \frac{(\widehat{x} + \widehat{\xi})^2 + (\sqrt{\widehat{x}^2 + 1} + \sqrt{\widehat{\xi}^2 + 1})^2}{(\widehat{x} - \widehat{\xi})^2 + (\sqrt{\widehat{x}^2 + 1} - \sqrt{\widehat{\xi}^2 + 1})^2} \cdot \frac{(\widehat{x} + \widehat{\xi})^2 + (\sqrt{\widehat{x}^2 + 1} - \sqrt{\widehat{\xi}^2 + 1})^2}{(\widehat{x} - \widehat{\xi})^2 + (\sqrt{\widehat{x}^2 + 1} + \sqrt{\widehat{\xi}^2 + 1})^2} \\ &= \frac{(\widehat{x} + \widehat{\xi})^2(1 + r_1^{-2})}{(\widehat{x} - \widehat{\xi})^2(1 + r_1^2)} \cdot \frac{(\widehat{x} + \widehat{\xi})^2 + (\widehat{x} - \widehat{\xi})^2 r_1^2}{(\widehat{x} - \widehat{\xi})^2 + (\widehat{x} + \widehat{\xi})^2 r_1^{-2}} = \frac{(\widehat{x} + \widehat{\xi})^2}{(\widehat{x} - \widehat{\xi})^2} \end{aligned} \tag{24}$$

after the cancelation.

Similarly, for K_1

$$\begin{aligned} & \frac{(\widehat{x} + \widehat{\xi})^2 + (\sqrt{\widehat{x}^2 + 1} + \sqrt{\widehat{\xi}^2 + 1})^2}{(\widehat{x} - \widehat{\xi})^2 + (\sqrt{\widehat{x}^2 + 1} - \sqrt{\widehat{\xi}^2 + 1})^2} \cdot \frac{(\widehat{x} - \widehat{\xi})^2 + (\sqrt{\widehat{x}^2 + 1} + \sqrt{\widehat{\xi}^2 + 1})^2}{(\widehat{x} + \widehat{\xi})^2 + (\sqrt{\widehat{x}^2 + 1} - \sqrt{\widehat{\xi}^2 + 1})^2} \\ &= \frac{(\widehat{x} + \widehat{\xi})^2}{(\widehat{x} - \widehat{\xi})^2} r_1^{-4}. \end{aligned} \tag{25}$$

Therefore, we have

$$K_2(x, \tau, y_\lambda) = K_2(x, \tau, y_0) \tag{26}$$

and

$$K_1(x, \tau, y_\lambda) = K_1(x, \tau, y_0) - 4 \log r_1. \tag{27}$$

Now, we are ready for the analysis of the asymptotics for the coefficients in L_λ .

1. The coefficient A_1 .

Consider $A_1(x, 0)$ first. We have

$$\begin{aligned} A_1(x, 0) &= \frac{1}{x \log^+(x^2)} \int_0^1 \log \left(\frac{x + \xi}{x - \xi} \right)^2 d\xi \\ &= \frac{1}{\log^+(x^2)} \int_0^{1/x} \log \left(\frac{1 + u}{1 - u} \right)^2 du = 2 + o(1), \quad x \rightarrow 0 \end{aligned} \tag{28}$$

and it is smooth in $(0, 1)$. At the point $x = 0$, we define $A_1(0, 0) = 2$, i.e., by its right limit.

Lemma 5.1. *We have*

$$\lim_{\lambda \rightarrow 0} \|A_1(x, \lambda) - A_1(x, 0)\|_{C[0,1]} = 0. \tag{29}$$

Proof. If $x = \lambda \widehat{x}$, then

$$\begin{aligned} & \int_0^1 K_1(x, \tau, y_\lambda) d\tau \\ &= \lambda \int_0^{1/\lambda} \log \left(\frac{(\widehat{x} + \widehat{\xi})^2 + (\sqrt{\widehat{x}^2 + 1} + \sqrt{\widehat{\xi}^2 + 1})^2}{(\widehat{x} - \widehat{\xi})^2 + (\sqrt{\widehat{x}^2 + 1} - \sqrt{\widehat{\xi}^2 + 1})^2} \cdot \frac{(\widehat{x} - \widehat{\xi})^2 + (\sqrt{\widehat{x}^2 + 1} + \sqrt{\widehat{\xi}^2 + 1})^2}{(\widehat{x} + \widehat{\xi})^2 + (\sqrt{\widehat{x}^2 + 1} - \sqrt{\widehat{\xi}^2 + 1})^2} \right) d\widehat{\xi}. \end{aligned}$$

Many estimates done below will be based on the following standard argument that we explain now in detail.

We have several regimes:

1. $\widehat{x} \in [0, 1]$. Notice that integration over any fixed interval $\widehat{\xi} \in [0, C]$ gives a contribution $O(\lambda)$, so we only need to control large $\widehat{\xi}$. Using (25), one gets the following asymptotics for the expression under the logarithm

$$\frac{(\widehat{x} + \widehat{\xi})^2}{(\widehat{x} - \widehat{\xi})^2} r_1^{-4} = \left(1 + \frac{4\widehat{x}}{\widehat{\xi}} + O(\widehat{\xi}^{-2}) \right) \left(1 + 4 \frac{\sqrt{\widehat{x}^2 + 1} - \widehat{x}}{\widehat{\xi}} + O(\widehat{\xi}^{-2}) \right), \quad \widehat{\xi} \rightarrow \infty.$$

Then, using the Taylor expansion for the logarithm, we get

$$\begin{aligned} & \lambda \int_0^{1/\lambda} \log \left(\frac{(\widehat{x} + \widehat{\xi})^2 + (\sqrt{\widehat{x}^2 + 1} + \sqrt{\widehat{\xi}^2 + 1})^2}{(\widehat{x} - \widehat{\xi})^2 + (\sqrt{\widehat{x}^2 + 1} - \sqrt{\widehat{\xi}^2 + 1})^2} \cdot \frac{(\widehat{x} - \widehat{\xi})^2 + (\sqrt{\widehat{x}^2 + 1} + \sqrt{\widehat{\xi}^2 + 1})^2}{(\widehat{x} + \widehat{\xi})^2 + (\sqrt{\widehat{x}^2 + 1} - \sqrt{\widehat{\xi}^2 + 1})^2} \right) d\widehat{\xi} \\ &= 4\lambda \sqrt{\widehat{x}^2 + 1} \log(1/\lambda) + O(\lambda) = 4\sqrt{x^2 + \lambda^2} \log(1/\lambda) + O(\lambda) \\ &= 2\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2) + O(\lambda). \end{aligned} \tag{30}$$

Given any fixed $\delta \in (0, 1)$, we have two cases.

2. Take $x \in (\delta, 1]$. We trivially get

$$\lim_{\lambda \rightarrow 0} \max_{x \in [\delta, 1]} \left| \int_0^1 K_1(x, \tau, y_\lambda) d\tau - \int_0^1 K_1(x, \tau, y_0) d\tau \right| = 0 \tag{31}$$

3. Let $x \in (\lambda, \delta]$. We substitute (23) to (25) and get

$$\begin{aligned} & \int_0^1 K_1(x, \tau, y_\lambda) d\tau \\ &= 2\lambda \int_0^{1/\lambda} \log \left| \frac{\widehat{x} + \widehat{\xi}}{\widehat{\xi} - \widehat{x}} \right| d\widehat{\xi} + 4\lambda \int_0^{1/\lambda} \log \left(1 + \frac{\sqrt{\widehat{x}^2 + 1} - \widehat{x}}{\widehat{x} + \widehat{\xi}} + O(\widehat{\xi}^{-2}) \right) d\widehat{\xi} \\ &= 2x \int_0^{1/x} \log \left| \frac{1+t}{1-t} \right| dt + 4\lambda(\sqrt{1 + \widehat{x}^2} - \widehat{x}) \int_0^{1/x} \frac{1}{1+t} dt + O(\lambda) \\ &= 4x \log(1/x) + O(x) + 4\lambda(\sqrt{1 + \widehat{x}^2} - \widehat{x}) \log(1/x) + O(\lambda) \\ &= 4 \log(1/x)(x + \lambda\sqrt{1 + \widehat{x}^2} - \lambda\widehat{x}) + O(x) \\ &= 2\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2) + O(x) \end{aligned} \tag{32}$$

The bounds above imply

$$\lim_{\lambda \rightarrow 0} \left\| \frac{\int_0^1 K_1(x, \tau, y_\lambda) d\tau}{\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} - \frac{\int_0^1 K_1(x, \tau, y_0) d\tau}{x \log^+(x^2)} \right\|_{L^\infty[0,1]} = 0. \tag{33}$$

Indeed, given any $\epsilon > 0$, we use (28), (30), and (32) to get

$$\left\| \frac{\int_0^1 K_1(x, \tau, y_\lambda) d\tau}{\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} - \frac{\int_0^1 K_1(x, \tau, y_0) d\tau}{x \log^+(x^2)} \right\|_{L^\infty[0,\delta]} \lesssim \frac{1}{\log^+(\delta^2 + \lambda^2)} < \epsilon/2$$

for $\delta < \delta(\epsilon)$ and $\lambda < \delta(\epsilon)$. For fixed $\delta < \delta(\epsilon)$, we have

$$\left\| \frac{\int_0^1 K_1(x, \tau, y_\lambda) d\tau}{\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} - \frac{\int_0^1 K_1(x, \tau, y_0) d\tau}{x \log^+(x^2)} \right\|_{L^\infty[\delta,1]} \leq \epsilon/2,$$

as long as $\lambda < \lambda(\epsilon)$ [by (31)]. This yields (33). \square

Later, we will need the following result

Lemma 5.2. *Suppose $\|g(x) - x\|_{Lip[0,1]} \leq \delta \ll 1$. Then,*

$$\left| \frac{\int_0^1 K_1(x, \tau, \sqrt{\lambda^2 + g^2(\tau)}) d\tau}{\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} \right| \lesssim 1$$

uniformly in $x \in [0, 1]$, $\lambda \in (0, 1]$, and g .

Its proof repeats the argument in the previous lemma (see also the proof of Lemma 7.2 below to check how the problem can be reduced to the homogeneous one for which the scaling can be easily performed to get the desired bound). This result can also be obtained by comparing to the case $g = x$ and using the stability estimates established in Lemma 7.1 below.

2. The coefficient A_2 .

Lemma 5.3. *For every fixed $\delta > 0$, we have*

$$A_2(x, \lambda) \rightarrow A_2(x, 0) = \frac{2 \log(x^{-2} + 1)}{x \log^+(x^2)}, \quad \lambda \rightarrow 0 \tag{34}$$

uniformly over $x \in [\delta, 1]$. Moreover, we have an estimate

$$A_2(x, \lambda) \sim \frac{1}{x} \tag{35}$$

which holds uniformly in $x \in (0, 1]$ and $\lambda \in (0, 1]$.

Proof. The expression for $A_2(x, 0)$ is easy to compute and the first part of the lemma is immediate. The formula for $A_2(x, \lambda)$ contains three terms. The first one involves K_1 and its asymptotics was established before. Consider the second term. By (21), we get

$$\int_0^1 y'_\lambda K_2(x, \tau, y_\lambda) d\tau = \lambda \int_0^{1/\lambda} \frac{\widehat{\xi}}{\sqrt{\widehat{\xi}^2 + 1}} \log \left(\frac{\widehat{x} + \widehat{\xi}}{\widehat{x} - \widehat{\xi}} \right)^2 d\widehat{\xi}.$$

The similar analysis yields:

(1) Uniformly in $x \in (\delta, 1]$, we get

$$\int_0^1 y'_\lambda K_2(x, \tau, y_\lambda) d\tau \rightarrow \int_0^1 K_2(x, \tau, y_0) d\tau, \quad \text{as } \lambda \rightarrow 0. \tag{36}$$

(2) If $\widehat{x} \in [0, 1]$, then we can split the integral into two. The first one is

$$\int_0^1 \frac{\widehat{\xi}}{\sqrt{\widehat{\xi}^2 + 1}} \log \left(\frac{\widehat{x} + \widehat{\xi}}{\widehat{x} - \widehat{\xi}} \right)^2 d\widehat{\xi}.$$

We have

$$\int_0^1 \widehat{\xi} \log \left(1 + \frac{2\widehat{x}\widehat{\xi}}{(\widehat{x} - \widehat{\xi})^2} \right) d\widehat{\xi} = \widehat{x}^2 \int_0^{\widehat{x}^{-1}} t \log \left(1 + \frac{2t}{(1-t)^2} \right) dt \sim \widehat{x}.$$

So, the integration over $[0, 1]$ amounts to $O(x)$ after multiplication by λ .

For the integral over $[1, \lambda^{-1}]$, we get

$$\int_1^{1/\lambda} \frac{\widehat{\xi}}{\sqrt{\widehat{\xi}^2 + 1}} \log \left(\frac{\widehat{x} + \widehat{\xi}}{\widehat{x} - \widehat{\xi}} \right)^2 d\widehat{\xi} = 4\widehat{x} \log(1/\lambda) + O(\widehat{x}).$$

Multiplication by λ yields

$$\int_0^1 y'_\lambda K_2(x, \tau, y_\lambda) d\tau = x \left(4 \log(1/\lambda) + O(1) \right).$$

(3) If $x \in (\lambda, \delta)$, then the integral over $[0, 1]$ can be handled as before and its contribution is at most \widehat{x}^{-1} . The integral over $[1, 1/\lambda]$ gives

$$\begin{aligned} \int_1^{1/\lambda} (1 + O(\widehat{\xi}^{-2})) \log \left(\frac{\widehat{x} + \widehat{\xi}}{\widehat{x} - \widehat{\xi}} \right)^2 d\widehat{\xi} &= \widehat{x} \int_{1/\widehat{x}}^{1/x} \left(1 + \frac{1}{\widehat{x}^2 t^2} \right) \log \left(\frac{1+t}{1-t} \right)^2 dt \\ &= 4\widehat{x} (\log(1/x) + O(1)) \end{aligned}$$

and we have

$$\int_0^1 y'_\lambda K_2(x, \tau, y_\lambda) d\tau = 4x (\log(1/x) + O(1)), \quad \lambda \rightarrow 0$$

Summarizing, we get the uniform bound

$$\int_0^1 y'_\lambda K_2(x, \tau, y_\lambda) d\tau = \begin{cases} 4x (\log(1/\lambda) + O(1)), & x < \lambda \\ 4x (\log(1/x) + O(1)), & x > \lambda \end{cases}. \tag{37}$$

For the third term in the expression for A_2 , we have

$$\begin{aligned}
 B_2 &= B_2^{(1)} + B_2^{(2)} \\
 B_2^{(1)} &= \frac{2x}{\sqrt{x^2 + \lambda^2}} \int_0^1 \left(x - \frac{\xi \sqrt{\lambda^2 + x^2}}{\sqrt{\lambda^2 + \xi^2}} \right) \left(\frac{y_\lambda(x) + y_\lambda(\xi)}{(x + \xi)^2 + (y_\lambda(x) + y_\lambda(\xi))^2} \right. \\
 &\quad \left. - \frac{y_\lambda(x) - y_\lambda(\xi)}{(x - \xi)^2 + (y_\lambda(x) - y_\lambda(\xi))^2} \right) d\xi \\
 B_2^{(2)} &= \frac{2x}{\sqrt{x^2 + \lambda^2}} \int_0^1 \left(x + \frac{\xi \sqrt{\lambda^2 + x^2}}{\sqrt{\lambda^2 + \xi^2}} \right) \left(\frac{y_\lambda(x) + y_\lambda(\xi)}{(x - \xi)^2 + (y_\lambda(x) + y_\lambda(\xi))^2} \right. \\
 &\quad \left. - \frac{y_\lambda(x) - y_\lambda(\xi)}{(x + \xi)^2 + (y_\lambda(x) - y_\lambda(\xi))^2} \right) d\xi
 \end{aligned}$$

Rescale the variables and recall the formulas (21) and (23).

One gets

$$B_2^{(1)} = -\lambda \frac{4\widehat{x}}{\sqrt{\widehat{x}^2 + 1}} \int_0^{1/\lambda} \frac{\widehat{\xi} r_1}{\sqrt{1 + \widehat{\xi}^2} (\widehat{x} \sqrt{1 + \widehat{\xi}^2} + \widehat{\xi} \sqrt{1 + \widehat{x}^2}) (1 + r_1^2)} d\widehat{\xi}. \tag{38}$$

As before, we consider two cases.

(1) $\widehat{x} \in [0, 1]$. For the integral over $[0, 1]$,

$$0 \leq \int_0^1 \frac{\widehat{\xi} r_1}{\sqrt{1 + \widehat{\xi}^2} (\widehat{x} \sqrt{1 + \widehat{\xi}^2} + \widehat{\xi} \sqrt{1 + \widehat{x}^2}) (1 + r_1^2)} d\widehat{\xi} \lesssim 1.$$

The other integral allows the estimate

$$\int_1^{1/\lambda} \frac{\widehat{\xi} r_1}{\sqrt{1 + \widehat{\xi}^2} (\widehat{x} \sqrt{1 + \widehat{\xi}^2} + \widehat{\xi} \sqrt{1 + \widehat{x}^2}) (1 + r_1^2)} d\widehat{\xi} \lesssim \log(1/\lambda)$$

since $r_1 \leq 1$.

(2) $\widehat{x} \in [1, 1/\lambda]$. We can write

$$0 \leq \int_0^{1/\lambda} \frac{\widehat{\xi} r_1}{\sqrt{1 + \widehat{\xi}^2} (\widehat{x} \sqrt{1 + \widehat{\xi}^2} + \widehat{\xi} \sqrt{1 + \widehat{x}^2}) (1 + r_1^2)} d\widehat{\xi} \lesssim \frac{\log(1/\lambda)}{\widehat{x}}. \tag{39}$$

For $B_2^{(2)}$, we have similarly

$$B_2^{(2)} = \frac{4\lambda\widehat{x}}{\sqrt{\widehat{x}^2 + 1}} \int_0^{1/\lambda} \frac{\widehat{x} \sqrt{\widehat{\xi}^2 + 1} + \widehat{\xi} \sqrt{\widehat{x}^2 + 1}}{\sqrt{\widehat{\xi}^2 + 1}} \cdot \frac{\widehat{\xi} r_1}{(\widehat{x} + \widehat{\xi})^2 + r_1^2 (\widehat{x} - \widehat{\xi})^2} d\widehat{\xi}. \tag{40}$$

(1) If $\widehat{x} \in [0, 1]$, we get

$$r_1 \lesssim \widehat{x} + \widehat{\xi}$$

and therefore

$$0 < \int_0^1 \frac{\widehat{x}\sqrt{\widehat{\xi}^2 + 1} + \widehat{\xi}\sqrt{\widehat{x}^2 + 1}}{\sqrt{\widehat{\xi}^2 + 1}} \cdot \frac{\widehat{\xi}r_1}{(\widehat{x} + \widehat{\xi})^2 + r_1^2(\widehat{x} - \widehat{\xi})^2} d\widehat{\xi} \lesssim 1.$$

For the other interval, we use $r_1 = 1 + O(\widehat{\xi}^{-1})$ to get

$$\begin{aligned} & \int_1^{1/\lambda} \frac{\widehat{x}\sqrt{\widehat{\xi}^2 + 1} + \widehat{\xi}\sqrt{\widehat{x}^2 + 1}}{\sqrt{\widehat{\xi}^2 + 1}} \cdot \frac{\widehat{\xi}r_1}{(\widehat{x} + \widehat{\xi})^2 + r_1^2(\widehat{x} - \widehat{\xi})^2} d\widehat{\xi} \\ &= \frac{\widehat{x} + \sqrt{\widehat{x}^2 + 1}}{2} \log(1/\lambda) + O(1). \end{aligned}$$

(2) If $\widehat{x} \in [1, 1/\lambda]$, then the asymptotics of r_1 yields

$$\begin{aligned} & \int_0^{1/\lambda} \frac{\widehat{x}\sqrt{\widehat{\xi}^2 + 1} + \widehat{\xi}\sqrt{\widehat{x}^2 + 1}}{\sqrt{\widehat{\xi}^2 + 1}} \cdot \frac{\widehat{\xi}r_1}{(\widehat{x} + \widehat{\xi})^2 + r_1^2(\widehat{x} - \widehat{\xi})^2} d\widehat{\xi} \\ & \sim \widehat{x} \int_0^{1/\lambda} \frac{\widehat{\xi}}{\widehat{x}^2 + \widehat{\xi}^2} d\widehat{\xi} \sim \widehat{x} \log^+ x. \end{aligned} \tag{41}$$

Now, the formulas (38) and (40) imply that $B_2^{(2)} \geq 0$ and $B_2^{(1)} \leq 0$. However, $B_2 = B_2^{(2)} + B_2^{(1)} \geq 0$. Indeed, this follows from (38), (40), and an estimate

$$\frac{\widehat{x}\sqrt{\widehat{\xi}^2 + 1} + \widehat{\xi}\sqrt{1 + \widehat{x}^2}}{(\widehat{x} + \widehat{\xi})^2 + r_1^2(\widehat{x} - \widehat{\xi})^2} \geq \frac{1}{(\widehat{x}\sqrt{\widehat{\xi}^2 + 1} + \widehat{\xi}\sqrt{1 + \widehat{x}^2})(1 + r_1^2)}.$$

Thus, we have

$$0 \leq B_2 \leq B_2^{(2)} \lesssim x \log^+ \lambda, \quad 0 < x < \lambda$$

and

$$0 \leq B_2 \leq B_2^{(2)} \lesssim x \log^+ x, \quad \lambda < x < 1.$$

Moreover, (39) and (41) provide a lower bound

$$B_2 \geq \lambda(C_1 \widehat{x} \log^+ x - C_2 \widehat{x}^{-1} \log^+ \lambda), \quad x > \lambda$$

and therefore

$$B_2 \geq C_3 x \log^+ x \tag{42}$$

for $\widehat{x} > C_4$ where C_4 is sufficiently large absolute constant.

Consider the sum of the first two terms in (20). We have

$$\begin{aligned} & \int_0^1 K_1(x, \tau, y_\lambda) d\tau - \frac{x}{\sqrt{\lambda^2 + x^2}} \int_0^1 y'_\lambda(\tau) K_2(x, \tau, y_\lambda) d\tau \\ &= 4 \log(1/\lambda)(\sqrt{x^2 + \lambda^2} - x) + O(\lambda), \quad 0 < x < \lambda \end{aligned} \tag{43}$$

and

$$= 4 \log(1/x)(\sqrt{x^2 + \lambda^2} - x) + O(x), \quad \lambda < x < \delta. \tag{44}$$

Add B_2 to this expression and divide by $x\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)$. On the interval $x \in (0, C_4\lambda)$, we use (43) and $B_2 \geq 0$ to get $A_2 \sim x^{-1}$. For $x \in (C_4\lambda, \delta)$, we apply (44) and (42) to produce that same bound. If $x \in [\delta, 1]$, we have convergence to $A_2(x, 0)$ which is positive. \square

Similar to Lemma 5.2, we have

Lemma 5.4. *Suppose $\|g(x) - x\|_{Lip[0,1]} \leq \delta \ll 1$. Then,*

$$\left| \frac{\int_0^1 \left(\sqrt{\lambda^2 + g^2(\tau)}\right)' K_2(x, \tau, \sqrt{\lambda^2 + g^2(\tau)}) d\tau}{x \log^+(x^2 + \lambda^2)} \right| \lesssim 1$$

uniformly in $x \in (0, 1]$, $\lambda \in (0, 1]$, and g .

This result can be proved directly or by comparison to the case when $g = x$ if the stability estimates (see (68) below) are used.

3. The Kernel $M(x, \xi, \lambda)$ and the Corresponding Operator

In this subsection, we will show that $M(x, \xi, \lambda) \rightarrow M(x, \xi, 0)$ in a suitable sense when $\lambda \rightarrow 0$. Recall that \mathcal{M}_λ is the integral operator with the kernel $M(x, \xi, \lambda)$. Below, we need to use the smooth cut-off ω_δ which was defined as follows. Let $\omega(x)$ be a smooth function such that $\omega(x) = 1$ on $|x| < 1/2$, $\omega(x) = 0$ on $|x| > 1$ and $0 \leq \omega(x) \leq 1$. For a parameter $\delta > 0$, we consider $\omega_\delta(x) = \omega(x/\delta)$ and $\omega_\delta^c(x) = 1 - \omega_\delta(x)$.

We have

Lemma 5.5. *Fix any $\delta > 0$. Then,*

$$\lim_{\lambda \rightarrow 0} \sup_{x > \delta} \int_0^1 |M(x, \xi, \lambda) - M(x, \xi, 0)| d\xi = 0$$

and therefore

$$\lim_{\lambda \rightarrow 0} \|\omega_\delta^c(x)(\mathcal{M}_\lambda - \mathcal{M}_0)\|_{L^\infty[0,1], L^\infty[0,1]} = 0.$$

Proof. We start with

$$\lim_{\lambda \rightarrow 0} \sup_{x > \delta} \int_0^1 |D_2(x, \xi, \lambda) - D_2(x, \xi, 0)| d\xi = 0.$$

By (26),

$$\int_0^1 |D_2(x, \xi, \lambda) - D_2(x, \xi, 0)| d\xi < C(\delta) \int_0^1 \left(1 - \frac{\xi}{\sqrt{\xi^2 + \lambda^2}}\right) \log \left| \frac{x + \xi}{x - \xi} \right| d\xi$$

and the last expression tends to zero uniformly in $x \in [\delta, 1]$ when $\lambda \rightarrow 0$.

To handle D_1 , we only need to show that

$$\lim_{\lambda \rightarrow 0} \sup_{x \in [\delta, 1], \xi \in [0, 1]} \left| \int_{\xi}^1 D_1(x, \tau, \lambda) d\tau - \int_{\xi}^1 D_1(x, \tau, 0) d\tau \right| = 0. \tag{45}$$

To this end, we first simplify the expression for $D_1(x, \tau, \lambda)$ using the formulas (21) and (23).

$$D_1(x, \xi, \lambda) = D_1^{(1)} + D_1^{(2)} + D_1^{(3)}, \tag{46}$$

where (below $x = \lambda \widehat{x}$ and $\xi = \lambda \widehat{\xi}$)

$$D_1^{(1)}(x, \xi, \lambda) = \frac{1}{x \sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} \cdot \frac{4 \widehat{\xi} \widehat{x} r_1}{(1 + r_1^2)(\widehat{\xi} \sqrt{1 + \widehat{x}^2} + \widehat{x} \sqrt{1 + \widehat{\xi}^2})(1 + \widehat{\xi}^2)}$$

$$D_1^{(2)}(x, \xi, \lambda) = \frac{1}{x \sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} \cdot \left(\frac{\widehat{x} \sqrt{1 + \widehat{\xi}^2} + \widehat{\xi} \sqrt{1 + \widehat{x}^2}}{1 + \widehat{\xi}^2} \right) \cdot \left(\frac{4 \widehat{x} \widehat{\xi} r_1}{(\widehat{x} + \widehat{\xi})^2 + (\widehat{x} - \widehat{\xi})^2 r_1^2} \right)$$

$$D_1^{(3)}(x, \xi, \lambda) = -\frac{1}{x \log^+(x^2 + \lambda^2)} \cdot \frac{\lambda^2}{(\lambda^2 + \xi^2)^{3/2}} \log \left(\frac{x + \xi}{x - \xi} \right)^2.$$

Since $D_1^{(3)}(x, \xi, 0) = 0$, we first show that

$$\sup_{x > \delta} \int_0^1 |D_1^{(3)}(x, \xi, \lambda)| d\xi \rightarrow 0, \quad \lambda \rightarrow 0.$$

To see that, first split the integral

$$\int_0^1 \frac{\lambda^2}{(\lambda^2 + \xi^2)^{3/2}} \log \left(\frac{x + \xi}{x - \xi} \right)^2 d\xi = \int_0^{\delta/2} \frac{\lambda^2}{(\lambda^2 + \xi^2)^{3/2}} \log \left(\frac{x + \xi}{x - \xi} \right)^2 d\xi + \int_{\delta/2}^1 \frac{\lambda^2}{(\lambda^2 + \xi^2)^{3/2}} \log \left(\frac{x + \xi}{x - \xi} \right)^2 d\xi.$$

The second integral goes to zero as $\lambda \rightarrow 0$ uniformly in $x > \delta$. The first one is bounded by

$$C \int_0^{\delta/2} \frac{\xi \lambda^2}{(\lambda^2 + \xi^2)^{3/2}} d\xi \lesssim \lambda.$$

All constants involved are δ dependent.

Similarly, $D_1^{(1)}(x, \xi, 0) = 0$ and we have

$$\sup_{x > \delta} \int_0^1 D_1^{(1)}(x, \xi, \lambda) d\xi \lesssim \lambda + \lambda \int_1^\infty \frac{\widehat{x} \widehat{\xi} d\widehat{\xi}}{(\widehat{x} \widehat{\xi})(1 + \widehat{\xi}^2)} \lesssim \lambda.$$

For $D_1^{(2)}(x, \xi, 0)$, we have

$$D_1^{(2)}(x, \xi, 0) = \frac{1}{x^2 \log^+(x^2)} \cdot \frac{4x^2}{x^2 + \xi^2}.$$

To show that

$$\lim_{\lambda \rightarrow 0} \sup_{x > \delta, \xi > 0} \int_{\xi}^1 |D_2^{(2)}(x, \tau, \lambda) - D_2^{(2)}(x, \tau, 0)| d\tau = 0$$

it is sufficient to prove

$$\lim_{\lambda \rightarrow 0} \sup_{x > \delta} \lambda \int_0^{1/\lambda} \left| \left(\frac{\widehat{x}\sqrt{1+\widehat{\xi}^2} + \widehat{\xi}\sqrt{1+\widehat{x}^2}}{1+\widehat{\xi}^2} \right) \cdot \left(\frac{4\widehat{x}\widehat{\xi}r_1}{(\widehat{x}+\widehat{\xi})^2 + (\widehat{x}-\widehat{\xi})^2 r_1^2} \right) - \frac{4\widehat{x}^2}{\widehat{x}^2 + \widehat{\xi}^2} \right| d\widehat{\xi} = 0.$$

The integral over any interval $[0, T]$ is uniformly bounded. For large \widehat{x} and $\widehat{\xi}$, we substitute

$$r_1 = 1 + O\left(\frac{1}{\widehat{x}\widehat{\xi}}\right), \quad \sqrt{1+\widehat{\xi}^2} = \widehat{\xi} + O(\widehat{\xi}^{-1}), \quad \sqrt{1+\widehat{x}^2} = \widehat{x} + O(\widehat{x}^{-1}).$$

Collecting the errors produced by this substitution, we estimate this expression by

$$\lambda \int_1^{1/\lambda} \frac{\widehat{x}^2}{\widehat{x}^2 + \widehat{\xi}^2} (\widehat{\xi}^{-2} + \widehat{x}^{-2}) d\widehat{\xi} \lesssim \lambda.$$

□

The next step is to estimate

$$\|\omega_\delta(x)\mathcal{M}_\lambda\|_{L^\infty[0,1], L^\infty[0,1]},$$

where δ and λ are small.

Lemma 5.6. *We have*

$$\lim_{\delta \rightarrow 0, \lambda \rightarrow 0} \|\omega_\delta(x)\mathcal{M}_\lambda\|_{L^\infty[0,1], L^\infty[0,1]} = 0. \tag{47}$$

Proof. We only need to show that

$$\lim_{\delta \rightarrow 0, \lambda \rightarrow 0} \sup_{x \in [0, \delta]} \int_0^1 \left| D_2(x, \xi, \lambda) + \int_{\xi}^1 D_1(x, \tau, \lambda) d\tau \right| d\xi = 0. \tag{48}$$

It is instructive to first do that calculation for $\lambda = 0$. In this case,

$$\begin{aligned} & \frac{1}{x \log^+ x} \int_0^1 \left| \left(2 \log \left| \frac{x+\xi}{x-\xi} \right| - \int_{\xi}^1 \frac{4x}{x^2 + \tau^2} d\tau \right) \right| d\xi \\ &= \frac{2}{\log^+ x} \int_0^{1/x} \left| \left(\log \left| \frac{1+\xi}{1-\xi} \right| - \int_{\xi}^{1/x} \frac{2}{1+\tau^2} d\tau \right) \right| d\xi. \end{aligned}$$

We have

$$\int_{\xi}^{1/x} \frac{2}{1+\tau^2} d\tau = \int_{\xi}^{\infty} \frac{2}{1+\tau^2} d\tau + O(x) = \frac{2}{\xi} + O(\xi^{-2} + x), \quad \xi \gg 1$$

and

$$\log \left| \frac{1+\xi}{1-\xi} \right| = \frac{2}{\xi} + O(\xi^{-2}).$$

This entails the necessary cancelation and a bound

$$\frac{1}{x \log^+ x} \int_0^1 \left| \left(2 \log \left| \frac{x + \xi}{x - \xi} \right| - \int_{\xi}^1 \frac{4x}{x^2 + \tau^2} d\tau \right) \right| d\xi \lesssim \frac{1}{\log^+ x}.$$

The logarithm in the denominator will give convergence to zero when $x \rightarrow 0$.

Now, we will need to prove analogous inequalities uniformly in small λ . The expression

$$\int_0^1 \left| D_2(x, \xi, \lambda) + \int_{\xi}^1 D_1(x, \tau, \lambda) d\tau \right| d\xi$$

will be handled term by term.

We start by proving

$$\lim_{\delta \rightarrow 0, \lambda \rightarrow 0} \sup_{x \in [0, \delta]} \int_0^1 \int_{\xi}^1 |D_1^{(3)}(x, \tau, \lambda)| d\tau d\xi = 0. \tag{49}$$

The integral is bounded by

$$\begin{aligned} & \frac{1}{\widehat{x} \log^+(\lambda^2 \widehat{x}^2 + \lambda^2)} \int_0^{1/\lambda} \int_{\widehat{\xi}}^{1/\lambda} \frac{1}{1 + \widehat{\tau}^3} \log \left| \frac{\widehat{x} + \widehat{\tau}}{\widehat{x} - \widehat{\tau}} \right| d\widehat{\tau} d\widehat{\xi} \\ &= \frac{1}{\widehat{x} \log^+(\lambda^2 \widehat{x}^2 + \lambda^2)} \int_0^{1/\lambda} \frac{\widehat{\tau}}{1 + \widehat{\tau}^3} \log \left| \frac{\widehat{x} + \widehat{\tau}}{\widehat{x} - \widehat{\tau}} \right| d\widehat{\tau} \end{aligned}$$

For the integral, an estimate holds

$$\int_0^{1/\lambda} \frac{\widehat{\tau}}{1 + \widehat{\tau}^3} \log \left| \frac{\widehat{x} + \widehat{\tau}}{\widehat{x} - \widehat{\tau}} \right| d\widehat{\tau} \lesssim \widehat{x} + \int_{1/\widehat{x}}^{1/\lambda} \widehat{x}^{-1} u^{-2} \log \left| \frac{1 + u}{1 - u} \right| du.$$

The last integral is bounded by $C\widehat{x}$ for $\widehat{x} < 1$. For $\widehat{x} > 1$, it is estimated by $C \frac{\log^+ \widehat{x}}{\widehat{x}}$. Since

$$\lim_{\lambda \rightarrow 0} \sup_{\widehat{x} \in (0, 1)} \frac{\widehat{x}}{\widehat{x} \log^+(\lambda^2 \widehat{x}^2 + \lambda^2)} \lesssim \lim_{\lambda \rightarrow 0} \frac{1}{\log^+ \lambda} = 0$$

and

$$\lim_{\lambda \rightarrow 0} \sup_{\widehat{x} > 1} \frac{\log^+ \widehat{x}}{\widehat{x}^2 \log^+(\lambda^2 \widehat{x}^2 + \lambda^2)} \lesssim \lim_{\lambda \rightarrow 0} \frac{1}{\log^+ \lambda} = 0$$

we get (49).

Consider the other terms

$$\begin{aligned} & \int_0^1 \left| D_2(x, \xi, \lambda) + \int_{\xi}^1 \left(D_1^{(1)}(x, \tau, \lambda) + D_1^{(2)}(x, \tau, \lambda) \right) d\tau \right| d\xi \\ & \lesssim \frac{1}{\widehat{x} \sqrt{\widehat{x}^2 + 1} \log^+(\lambda^2 \widehat{x}^2 + \lambda^2)} \left(\int_0^{1/\lambda} \left| \frac{\sqrt{\widehat{x}^2 + 1}}{\sqrt{\widehat{\xi}^2 + 1}} \widehat{\xi} \log \left(\frac{\widehat{x} + \widehat{\xi}}{\widehat{x} - \widehat{\xi}} \right) \right|^2 \right. \\ & \quad - \int_{\widehat{\xi}}^{1/\lambda} \left(\frac{4\widehat{\tau} \widehat{x} r_1}{(1 + r_1^2)(1 + \widehat{\tau}^2)(\widehat{\tau} \sqrt{1 + \widehat{x}^2} + \widehat{x} \sqrt{1 + \widehat{\tau}^2})} \right. \\ & \quad \left. \left. + \frac{\widehat{x} \sqrt{1 + \widehat{\tau}^2} + \widehat{\tau} \sqrt{1 + \widehat{x}^2}}{1 + \widehat{\tau}^2} \cdot \frac{4\widehat{x} \widehat{\tau} r_1}{(\widehat{x} + \widehat{\tau})^2 + (\widehat{x} - \widehat{\tau})^2 r_1^2} \right) d\widehat{\tau} \right| d\widehat{\xi} \Big). \end{aligned}$$

We consider two cases.

(1) Take $\widehat{x} \in (0, 1]$. First, let $\widehat{\xi} \in (0, 1)$. We get

$$\begin{aligned} & \int_0^1 \left| \frac{\sqrt{\widehat{x}^2 + 1}}{\sqrt{\widehat{\xi}^2 + 1}} \widehat{\xi} \log \left(\frac{\widehat{x} + \widehat{\xi}}{\widehat{x} - \widehat{\xi}} \right)^2 - \int_{\widehat{\xi}}^{1/\lambda} \left(\frac{4\widehat{\tau}\widehat{x}r_1}{(1+r_1^2)(1+\widehat{\tau}^2)(\widehat{\tau}\sqrt{1+\widehat{x}^2} + \widehat{x}\sqrt{1+\widehat{\tau}^2})} \right. \right. \\ & \quad \left. \left. + \frac{\widehat{x}\sqrt{1+\widehat{\tau}^2} + \widehat{\tau}\sqrt{1+\widehat{x}^2}}{1+\widehat{\tau}^2} \cdot \frac{4\widehat{x}\widehat{\tau}r_1}{(\widehat{x}+\widehat{\tau})^2 + (\widehat{x}-\widehat{\tau})^2r_1^2} \right) d\widehat{\tau} \right| d\widehat{\xi} \\ & \lesssim \widehat{x} + \widehat{x} \int_0^1 d\widehat{\xi} \int_{\widehat{\xi}}^{1/\lambda} \left(\frac{\widehat{\tau}}{(1+\widehat{\tau}^2)(\widehat{\tau} + \widehat{x} + \widehat{\tau}\widehat{x})} + \frac{\widehat{\tau}(\widehat{\tau} + \widehat{x} + \widehat{\tau}\widehat{x})}{(1+\widehat{\tau}^2)(\widehat{x}^2 + \widehat{\tau}^2)} \right) d\widehat{\tau} \lesssim \widehat{x}. \end{aligned}$$

Thus, this gives $O((\log^+ \lambda)^{-1})$ contribution when divided by $\widehat{x}\sqrt{\widehat{x}^2 + 1} \log^+(\lambda^2\widehat{x}^2 + \lambda^2)$. If $\widehat{\xi} > 1$, we can use the asymptotical formulas $r_1 = 1 + O(\widehat{\xi}^{-1})$ and $\sqrt{\widehat{\xi}^2 + 1} = \widehat{\xi} + O(\widehat{\xi}^{-1})$ to get

$$\begin{aligned} & \int_1^{1/\lambda} \left| \frac{\sqrt{\widehat{x}^2 + 1}}{\sqrt{\widehat{\xi}^2 + 1}} \widehat{\xi} \log \left(\frac{\widehat{x} + \widehat{\xi}}{\widehat{x} - \widehat{\xi}} \right)^2 - \int_{\widehat{\xi}}^{1/\lambda} \left(\frac{4\widehat{\tau}\widehat{x}r_1}{(1+r_1^2)(1+\widehat{\tau}^2)(\widehat{\tau}\sqrt{1+\widehat{x}^2} + \widehat{x}\sqrt{1+\widehat{\tau}^2})} \right. \right. \\ & \quad \left. \left. + \frac{\widehat{x}\sqrt{1+\widehat{\tau}^2} + \widehat{\tau}\sqrt{1+\widehat{x}^2}}{1+\widehat{\tau}^2} \cdot \frac{4\widehat{x}\widehat{\tau}r_1}{(\widehat{x}+\widehat{\tau})^2 + (\widehat{x}-\widehat{\tau})^2r_1^2} \right) d\widehat{\tau} \right| d\widehat{\xi} \\ & = \int_1^{1/\lambda} \left| \frac{4\widehat{x}\sqrt{\widehat{x}^2 + 1}}{\widehat{\xi}} (1 + O(\widehat{\xi}^{-2} + \widehat{x}\widehat{\xi}^{-1})) \right. \\ & \quad \left. - \int_{\widehat{\xi}}^{1/\lambda} \left(\frac{2\widehat{x}}{\widehat{x} + \sqrt{\widehat{x}^2 + 1}} + 2\widehat{x}(\widehat{x} + \sqrt{1 + \widehat{x}^2}) \right) \widehat{\tau}^{-2} + \widehat{x}O(\widehat{\tau}^{-3}) d\widehat{\tau} \right| d\widehat{\xi} \lesssim \widehat{x}. \end{aligned}$$

Indeed,

$$\frac{2\widehat{x}}{\widehat{x} + \sqrt{\widehat{x}^2 + 1}} + 2\widehat{x}(\widehat{x} + \sqrt{1 + \widehat{x}^2}) = 4\widehat{x}\sqrt{1 + \widehat{x}^2}$$

and we have cancelation of the main terms.

Summing up these estimates, we get

$$\int_0^1 \left| D_2(x, \xi, \lambda) + \int_{\xi}^1 (D_1^{(1)}(x, \tau, \lambda) + D_1^{(2)}(x, \tau, \lambda)) d\tau \right| d\xi \lesssim \frac{1}{\log^+ \lambda}, \quad x \in (0, \lambda). \tag{50}$$

(2) Consider the case when $\widehat{x} > 1$. First, take $\widehat{\xi} \in (0, 1)$. We get

$$\int_0^1 \left| \frac{\sqrt{\widehat{x}^2 + 1}}{\sqrt{\widehat{\xi}^2 + 1}} \widehat{\xi} \log \left(\frac{\widehat{x} + \widehat{\xi}}{\widehat{x} - \widehat{\xi}} \right)^2 \right| d\widehat{\xi} \lesssim 1$$

and

$$\begin{aligned} & \int_0^1 \int_{\widehat{\xi}}^{1/\lambda} \left(\frac{2\widehat{\tau}\widehat{x}r_1}{(1+r_1^2)(1+\widehat{\tau}^2)(\widehat{\tau}\sqrt{1+\widehat{x}^2} + \widehat{x}\sqrt{1+\widehat{\tau}^2})} \right. \\ & \quad \left. + \frac{\widehat{x}\sqrt{1+\widehat{\tau}^2} + \widehat{\tau}\sqrt{1+\widehat{x}^2}}{1+\widehat{\tau}^2} \cdot \frac{4\widehat{x}\widehat{\tau}r_1}{(\widehat{x}+\widehat{\tau})^2 + (\widehat{x}-\widehat{\tau})^2r_1^2} \right) d\widehat{\tau} d\widehat{\xi} \lesssim 1 + \widehat{x}. \end{aligned}$$

Thus, this gives the contribution bounded by

$$\sup_{\widehat{x} > 1} \frac{1}{\widehat{x} \log^+(\lambda^2 \widehat{x}^2 + \lambda^2)} \lesssim \frac{1}{\log^+ \lambda}.$$

For the interval $\widehat{\xi} \in (1, \lambda^{-1})$, we again use asymptotics for $r_1, \sqrt{\widehat{x}^2 + 1}$, and $\sqrt{\widehat{\xi}^2 + 1}$:

$$\int_1^{1/\lambda} \left| \sqrt{\widehat{x}^2 + 1} (1 + O(\widehat{\xi}^{-2})) \log \left(\frac{\widehat{x} + \widehat{\xi}}{\widehat{x} - \widehat{\xi}} \right)^2 - \int_{\widehat{\xi}}^{1/\lambda} \frac{2\widehat{x}}{\widehat{\tau}} \left(\frac{1}{\widehat{\tau} \sqrt{1 + \widehat{x}^2} + \widehat{x}\widehat{\tau}} + \frac{\widehat{\tau} \sqrt{1 + \widehat{x}^2} + \widehat{x}\widehat{\tau}}{\widehat{x}^2 + \widehat{\tau}^2} \right) (1 + O(\widehat{\tau}^{-1} \widehat{x}^{-1} + \widehat{\tau}^{-2})) d\widehat{\tau} \right| d\widehat{\xi}.$$

The errors produce the term bounded by $C(\log^+ \widehat{x} + \widehat{x})$ and the change of variables in the integrals gives

$$\int_{1/\widehat{x}}^{1/x} \left| \widehat{x} \sqrt{1 + \widehat{x}^2} \log \left(\frac{1 + u}{1 - u} \right)^2 - \int_u^{1/x} 2\widehat{x}\widehat{\tau}^{-1} \left(\frac{1}{\widehat{\tau}(\sqrt{\widehat{x}^2 + 1} + \widehat{x})} + \frac{\widehat{\tau}(\sqrt{\widehat{x}^2 + 1} + \widehat{x})}{\widehat{\tau}^2 + 1} \right) d\widehat{\tau} \right| du$$

First, notice that

$$\widehat{x} \int_{1/\widehat{x}}^{1/x} \left| \int_{1/x}^{\infty} 2\widehat{\tau}^{-1} \left(\frac{1}{\widehat{\tau}(\sqrt{\widehat{x}^2 + 1} + \widehat{x})} + \frac{\widehat{\tau}(\sqrt{\widehat{x}^2 + 1} + \widehat{x})}{\widehat{\tau}^2 + 1} \right) d\widehat{\tau} \right| du \lesssim \widehat{x}^2.$$

Then,

$$\int_{1/\widehat{x}}^1 \left| \widehat{x} \sqrt{1 + \widehat{x}^2} \log \left(\frac{1 + u}{1 - u} \right)^2 - \widehat{x} \int_u^{\infty} 2\widehat{\tau}^{-1} \left(\frac{1}{\widehat{\tau}(\sqrt{\widehat{x}^2 + 1} + \widehat{x})} + \frac{\widehat{\tau}(\sqrt{\widehat{x}^2 + 1} + \widehat{x})}{\widehat{\tau}^2 + 1} \right) d\widehat{\tau} \right| du \lesssim \widehat{x}^2 + \log^+ \widehat{x}$$

and

$$\int_1^{1/x} \left| \widehat{x} \sqrt{1 + \widehat{x}^2} \log \left(\frac{1 + u}{1 - u} \right)^2 - \int_u^{\infty} \frac{2\widehat{x}}{\widehat{\tau}} \left(\frac{1}{\widehat{\tau}(\sqrt{\widehat{x}^2 + 1} + \widehat{x})} + \frac{\widehat{\tau}(\sqrt{\widehat{x}^2 + 1} + \widehat{x})}{\widehat{\tau}^2 + 1} \right) d\widehat{\tau} \right| du \lesssim \widehat{x}^2$$

after the cancelation of the main terms in the asymptotics. Collecting these bounds, we get

$$\int_0^1 \left| D_2(x, \xi, \lambda) + \int_{\xi}^1 (D_1^{(1)}(x, \tau, \lambda) + D_1^{(2)}(x, \tau, \lambda)) d\tau \right| d\xi \leq \frac{\widehat{x}}{\sqrt{\widehat{x}^2 + 1} \log^+(x^2 + \lambda^2)}$$

which [together with (49) and (50)] gives (48) and finishes the proof. \square

We immediately get the following

Corollary 5.1.

$$\lim_{\lambda \rightarrow 0} \|\mathcal{M}_\lambda - \mathcal{M}_0\|_{L^\infty[0,1], L^\infty[0,1]} = 0$$

Proof. It is sufficient to apply Lemma 5.5 and Lemma 5.6. \square

5.3. *Inverting L_λ .* Divide the equation

$$L_\lambda u = g$$

by $A_1(x, \lambda)$ to rewrite it as

$$u' + pu + \int_0^1 M_2(x, \xi, \lambda)u'(\xi)d\xi = g_1$$

where

$$p(x, \lambda) = \frac{A_2(x, \lambda)}{A_1(x, \lambda)}$$

and

$$M_2(x, \xi, \lambda) = \frac{M(x, \xi, \lambda)}{A_1(x, \lambda)}, \quad g_1 = \frac{g(x)}{A_1(x, \lambda)}.$$

Due to (28) and (29), this is a minor change as far as inversion of L_λ is concerned.

The equation

$$u' + pu = F, \quad u(0) = 0$$

has the solution

$$u = \int_0^x \exp\left(-\int_\xi^x p(t)dt\right) F(\xi)d\xi$$

and therefore

$$u' = F - p \int_0^x \exp\left(-\int_\xi^x p(t)dt\right) F(\xi)d\xi.$$

This is the same as

$$\begin{aligned} u'(x) &= g_2(x) - \int_0^1 M_2(x, \xi, \lambda)u'(\xi)d\xi \\ &+ p(x) \int_0^x \exp\left(-\int_t^x p(\tau)d\tau\right) \int_0^1 M_2(t, \xi, \lambda)u'(\xi)d\xi dt \end{aligned} \quad (51)$$

and

$$g_2(x) = g_1(x) - p(x) \int_0^x \exp\left(-\int_\xi^x p(t)dt\right) g_1(\xi)d\xi.$$

We can rewrite

$$u' + O_\lambda u' = B_\lambda g, \quad u' = (I + O_\lambda)^{-1} B_\lambda g \quad (52)$$

provided that $I + O_\lambda$ is invertible. The expressions for O_λ and B_λ are as follows

$$B_\lambda g = g_2 = \frac{g(x)}{A_1(x, \lambda)} - \frac{A_2(x, \lambda)}{A_1(x, \lambda)} \int_0^x \exp\left(-\int_\xi^x \frac{A_2(t, \lambda)}{A_1(t, \lambda)} dt\right) \frac{g(\xi)}{A_1(\xi, \lambda)} d\xi$$

and

$$O_\lambda f = \frac{1}{A_1(x, \lambda)} (\mathcal{M}_\lambda f)(x) - \frac{A_2(x, \lambda)}{A_1(x, \lambda)} \int_0^x \exp\left(-\int_\xi^x \frac{A_2(t, \lambda)}{A_1(t, \lambda)} dt\right) \frac{(\mathcal{M}_\lambda f)(\xi)}{A_1(\xi, \lambda)} d\xi. \tag{53}$$

Lemma 5.7. *We have*

$$\|B_\lambda\|_{L^\infty[0,1], L^\infty[0,1]} \lesssim 1 \tag{54}$$

uniformly in $\lambda \in (0, \lambda_0)$.

Proof. Since both A_1 and A_2 are positive, we have

$$|B_\lambda g(x)| \leq C \left(\|g\|_{L^\infty[0,1]} + \frac{1}{x} \int_0^x |g(\xi)| d\xi \right)$$

uniformly in $\lambda \in (0, \lambda_0)$ and $x \in (0, 1]$ as follows from the analysis of A_1 and A_2 . This gives (54). \square

Consider O_λ . We have

Lemma 5.8.

$$\|O_\lambda - O_0\|_{L^\infty[0,1], L^\infty[0,1]} \rightarrow 0, \quad \lambda \rightarrow 0$$

Proof. For the first term,

$$\left\| \frac{1}{A_1(x, \lambda)} \mathcal{M}_\lambda f - \frac{1}{A_1(x, 0)} \mathcal{M}_0 f \right\|_{L^\infty[0,1]} = o(1) \|f\|_{L^\infty[0,1]},$$

and $o(1) \rightarrow 0$ when $\lambda \rightarrow 0$, uniformly in f . Indeed, this follows from corollary 5.1 and the properties of $A_1(x, \lambda)$.

The second term can be written as

$$\begin{aligned} &\omega_\delta(x) \cdot \frac{A_2(x, \lambda)}{A_1(x, \lambda)} \int_0^x \exp\left(-\int_\xi^x \frac{A_2(t, \lambda)}{A_1(t, \lambda)} dt\right) \frac{(\mathcal{M}_\lambda f)(\xi)}{A_1(\xi, \lambda)} d\xi \\ &+ \omega_\delta^c(x) \cdot \frac{A_2(x, \lambda)}{A_1(x, \lambda)} \int_0^x \exp\left(-\int_\xi^x \frac{A_2(t, \lambda)}{A_1(t, \lambda)} dt\right) \frac{(\mathcal{M}_\lambda f)(\xi)}{A_1(\xi, \lambda)} d\xi \end{aligned}$$

where $\delta > 0$. If we denote the first/second expressions by $S_{1(2)}$, then

$$|S_1| \lesssim \frac{1}{x} \int_0^x \chi_{\xi < \delta} \left| \frac{(\mathcal{M}_\lambda f)(\xi)}{A_1(\xi, \lambda)} \right| d\xi$$

and

$$\lim_{\delta \rightarrow 0, \lambda \rightarrow 0} \|S_1\|_{L^\infty[0,1], L^\infty[0,1]} = 0$$

The last equality follows from (47).

For S_2 , one can write similarly

$$S_2 = \omega_\delta^c(x) \cdot \frac{A_2(x, \lambda)}{A_1(x, \lambda)} \int_0^x \chi_{\xi < \delta} \cdot \exp\left(-\int_\xi^x \frac{A_2(t, \lambda)}{A_1(t, \lambda)} dt\right) \frac{(\mathcal{M}_\lambda f)(\xi)}{A_1(\xi, \lambda)} d\xi \\ + \omega_\delta^c(x) \cdot \frac{A_2(x, \lambda)}{A_1(x, \lambda)} \int_0^x \chi_{\xi > \delta} \cdot \exp\left(-\int_\xi^x \frac{A_2(t, \lambda)}{A_1(t, \lambda)} dt\right) \frac{(\mathcal{M}_\lambda f)(\xi)}{A_1(\xi, \lambda)} d\xi.$$

The first expression can be handled in the same way. For the second, we consider

$$\left\| \omega_\delta^c(x) \cdot \frac{A_2(x, \lambda)}{A_1(x, \lambda)} \int_0^x \chi_{\xi > \delta} \cdot \exp\left(-\int_\xi^x \frac{A_2(t, \lambda)}{A_1(t, \lambda)} dt\right) \frac{(\mathcal{M}_\lambda f)(\xi)}{A_1(\xi, \lambda)} d\xi \right. \\ \left. - \omega_\delta^c(x) \cdot \frac{A_2(x, 0)}{A_1(x, 0)} \int_0^x \chi_{\xi > \delta} \cdot \exp\left(-\int_\xi^x \frac{A_2(t, 0)}{A_1(t, 0)} dt\right) \frac{(\mathcal{M}_0 f)(\xi)}{A_1(\xi, 0)} d\xi \right\|_{L^\infty[0,1]}.$$

If $\delta > 0$ is fixed, this expression is bounded by $o(1)\|f\|_{L^\infty[0,1]}$ as $\lambda \rightarrow 0$ (with constant depending on δ). That follows directly from the properties of $A_{1(2)}$ and \mathcal{M}_λ . Combining the obtained estimates we get the statement of the lemma. \square

By the standard argument of the perturbation theory, this lemma implies that inversion of $I + O_\lambda$ can be reduced to showing that $I + O_0$ is invertible. In the next section, we will check that.

5.3.1. The operator O_0 and its properties

Theorem 5.1. *The operator $I + O_0$ is invertible in $L^\infty[0, 1]$.*

Proof. For the case $\lambda = 0$, the formulas are very simple. We recall that

$$K_1(x, \xi, y_0) = K_2(x, \xi, y_0) = \log\left(\frac{x + \xi}{x - \xi}\right)^2.$$

Then, (28) and (34) imply that

$$A_1(x, 0) = \frac{1}{\log^+ x^2} \int_0^{1/x} \log\left(\frac{1 + \xi}{1 - \xi}\right)^2 d\xi = 2 + o(1), \quad x \rightarrow 0$$

and

$$A_2(x, 0) = \frac{2}{x \log^+ x^2} \log(x^{-2} + 1) = \frac{2}{x} + o(x), \quad x \rightarrow 0.$$

Thus,

$$p(x, 0) = \frac{A_2(x, 0)}{A_1(x, 0)} = \frac{1}{x} + o(x), \quad x \rightarrow 0.$$

Then,

$$D_2(x, \xi) = -\frac{1}{x \log^+ x^2} \log\left(\frac{x + \xi}{x - \xi}\right)^2$$

and

$$D_1(x, \xi) = \frac{4}{(x^2 + \xi^2) \log^+ x^2}.$$

Therefore,

$$M(x, \xi, 0) = \frac{1}{x \log^+ x^2} \left(-\log \left(\frac{x + \xi}{x - \xi} \right)^2 + \int_{\xi}^1 \frac{4\tau}{x^2 + \tau^2} d\tau \right)$$

and

$$M_2(x, \xi, 0) = \frac{M(x, \xi, 0)}{A_1(x, 0)}$$

where $A_1 \sim 1$ on all of $[0, 1]$.

Lemma 5.9. *The operator*

$$G_2 f = \int_0^1 M_2(x, \xi, 0) f(\xi) d\xi$$

is compact in $L^\infty[0, 1]$.

Proof. First, notice that

$$|G_2 f| \lesssim \frac{\|f\|_\infty}{\log^+ x} \int_0^{1/x} \left| \log \left(\frac{1 + \xi}{1 - \xi} \right)^2 - 4 \int_{\xi}^{1/x} \frac{d\tau}{\tau^2 + 1} \right| d\xi \lesssim \frac{\|f\|_\infty}{\log^+ x} \quad (55)$$

and thus G_2 is bounded in $L^\infty[0, 1]$.

The compactness now follows by the standard approximation argument. Let us write a partition of unity $1 = \phi_\delta + \phi_\delta^c$. Then, (55) yields $\|\phi_\delta G_2\|_{L^\infty[0,1], L^\infty[0,1]} \rightarrow 0$ as $\delta \rightarrow 0$. Then, for fixed $\delta > 0$, $\phi_\delta^c G_2$ is compact since the kernel has a weak singularity on the diagonal and is smooth away from it. Since the space of compact operators is closed in the operator topology, we have the statement of the lemma. \square

For O_0 , one gets

$$\begin{aligned} O_0 f &= \frac{1}{A_1(x, 0)} \mathcal{M}_0 f - \frac{A_2(x, 0)}{A_1(x, 0)} \int_0^x \exp \left(- \int_{\xi}^x \frac{A_2(t, 0)}{A_1(t, 0)} dt \right) \frac{(\mathcal{M}_0 f)(\xi)}{A_1(\xi, 0)} d\xi \\ &= (G_2 f)(x) - \frac{A_2(x, 0)}{A_1(x, 0)} \int_0^x \exp \left(- \int_{\xi}^x \frac{A_2(t, 0)}{A_1(t, 0)} dt \right) (G_2 f)(\xi) d\xi. \end{aligned} \quad (56)$$

Since the operator G_3 defined by

$$G_3 f = \frac{A_2(x, 0)}{A_1(x, 0)} \int_0^x \exp \left(- \int_{\xi}^x \frac{A_2(t, 0)}{A_1(t, 0)} dt \right) f(\xi) d\xi$$

is bounded in $L^\infty[0, 1]$, we get the compactness for O_0 in view of Lemma 5.9. Therefore, the Fredholm theory is applicable to $I + O_0$. In particular, to prove invertibility of $I + O_0$, we only need to check that its kernel is trivial.

Consider the equation

$$(I + O_0) f = 0$$

and suppose that $f \in L^\infty[0, 1]$. Recall (52). The equation

$$L_0 u = 0, \quad u \in \dot{L}ip[0, 1] \tag{57}$$

is equivalent to

$$(I + O_0)u' = 0, \quad u(x) = \int_0^x f(t)dt.$$

Thus, we only need to check that L_0 has zero kernel in $\dot{L}ip[0, 1]$.

The Eq. (57) is equivalent to

$$\int_0^1 (u'(x) - u'(\xi))K_1(x, \xi, y_0)d\xi + 8 \int_0^1 H'(2x^2 + 2\xi^2)(\xi u(x) + xu(\xi))d\xi = 0, \quad u \in \dot{L}ip[0, 1]$$

where

$$K_1(x, \xi, y_0) = H(2(x + \xi)^2) - H(2(x - \xi)^2) = \log \left(\frac{x + \xi}{x - \xi} \right)^2$$

since $H(x) = \log x$ in that case. Multiply the both sides by u and integrate over $[0, 1]$. For the general H , we have

$$\begin{aligned} &0.5 \int_0^1 (u(1) - u(\xi))^2 \left(H(2(1 + \xi)^2) - H(2(1 - \xi)^2) \right) d\xi \\ &- 2 \int_0^1 \int_0^1 (u(x) - u(\xi))^2 \left(H'(2(x + \xi)^2)(x + \xi) - H'(2(x - \xi)^2)(x - \xi) \right) dx d\xi \\ &+ 8 \int_0^1 u^2(x) \int_0^1 \xi H'(2x^2 + 2\xi^2) d\xi dx + 8 \int_0^1 \int_0^1 u(x)u(\xi)x H'(2\xi^2 + 2x^2) dx d\xi \\ &= \mathfrak{J}_1 + \dots + \mathfrak{J}_4. \end{aligned}$$

Let us study this expression term by term.

If $u_1(x) = u(1) - u(x)$, then

$$\mathfrak{J}_1 = \int_0^1 u_1^2(x) \log \left| \frac{1+x}{1-x} \right| dx \geq 0.$$

This is actually true for generic H that are monotonically increasing.

Using the symmetrization of the integrals, we get the following expressions

$$\begin{aligned} \mathfrak{J}_2 &= - \int_0^1 \int_0^1 \frac{(u(x) - u(\xi))^2}{x + \xi} dx d\xi \\ &= -2 \int_0^1 u^2(x) \log \left(\frac{1+x}{x} \right) dx + 2 \int_0^1 \int_0^1 \frac{u(x)u(\xi)}{x + \xi} dx d\xi \\ \mathfrak{J}_3 &= 2 \int_0^1 u^2(x) \log(1 + x^{-2}) dx \\ \mathfrak{J}_4 &= 4 \int_0^1 \int_0^1 u(x)u(\xi) \frac{x}{x^2 + \xi^2} dx d\xi = 2 \int_0^1 \int_0^1 u(x)u(\xi) \frac{x + \xi}{x^2 + \xi^2} dx d\xi \end{aligned}$$

Notice now that the sum of the first term in \mathfrak{J}_2 and \mathfrak{J}_3 is

$$2 \int_0^1 u^2(x) \log \left(\frac{x + x^{-1}}{1 + x} \right) dx \geq 0$$

because $x + x^{-1} \geq x + 1$ if $x \in (0, 1]$.

In the calculations that follow, the condition $u(x) = O(x)$, $x \rightarrow 0$ will ensure the convergence of all integrals involved. Since the Hilbert matrix is nonnegative ([12], proof of the theorem 5.3.1.), the integral

$$\mathcal{G}(u) = \int_0^1 \frac{u(\xi)}{x + \xi} d\xi$$

defines a positive definite operator in $L^2(0, 1)$. Thus,

$$\mathcal{G}_1(u) = \int_0^1 \frac{u(\xi)}{x^2 + \xi^2} d\xi$$

is positive definite as well, as the change of variables in the quadratic form shows. Also,

$$\frac{x + \xi}{x^2 + \xi^2} = \frac{1}{x + \xi} + \frac{2x\xi}{(x^2 + \xi^2)(x + \xi)}.$$

So, we only need to establish that

$$\mathcal{G}_2 u = \int_0^1 \frac{x\xi u(\xi)}{(x^2 + \xi^2)(x + \xi)} d\xi$$

is positive definite. That, however, is the corollary of the Schur’s theorem for the Hadamard product of the positive definite matrices ([12], p. 319), written for the integral operators (e.g., by the Riemann sum approximation). Indeed, it is sufficient to notice that

$$\frac{x\xi}{x^2 + \xi^2}$$

is a positive definite kernel (again, by the change of variables in the quadratic form). \square

Summing up the results of this section, we obtain (14).

6. $\|\psi_0\|_{\dot{L}ip[0,1]}$ is Small

In this section, we will prove (17), the smallness of initial data for the contraction mapping.

Lemma 6.1. *We have*

$$\|\psi_0\|_{\dot{L}ip[0,1]} = o(1), \quad \lambda \rightarrow 0. \tag{58}$$

Proof. As it follows from the previous section, we only need to show

$$\|F(x, \lambda)\|_{L^\infty[0,1]} = o(1), \quad \lambda \rightarrow 0. \tag{59}$$

Recall the definition of F ,

$$F(x, \lambda) = \frac{1}{x\sqrt{\lambda^2 + x^2} \log^+(x^2 + \lambda^2)} \times \left(x \int_0^1 K_1(x, \tau, y_\lambda) d\tau - \sqrt{\lambda^2 + x^2} \int_0^1 y'_\lambda(\tau) K_2(x, \tau, y_\lambda) d\tau \right)$$

For any given $\delta > 0$, we clearly have

$$\lim_{\lambda \rightarrow 0} \|\omega_\delta^c \cdot F(x, \lambda)\|_{L^\infty[0,1]} = 0.$$

For $x < \delta$, we can use the asymptotics established above [e.g., (30), (32), and (37)]. This gives

$$\|\omega_\delta \cdot F(x, \lambda)\|_{L^\infty[0,1]} \lesssim \frac{1}{\log^+(\delta^2 + \lambda^2)}.$$

These two estimates finish the proof of the lemma. \square

7. The Frechet Differentiability

In this section, we study $Q(u)$ given by

$$Q(u) = F(f, \lambda) - F(x, \lambda) - D_f F(x, \lambda)u, \quad f = x + u$$

and prove (15) and (16). We assume in this section that $\lambda \in (0, 1)$. Notice first that $Q(0) = 0$ and therefore (15) follows from (16). Let us prove (16).

We write

$$\begin{aligned} Q(u_2) &= F(x + u_2, \lambda) - F(x, \lambda) - D_f F(x, \lambda)(x + u_2) \\ Q(u_1) &= F(x + u_1, \lambda) - F(x, \lambda) - D_f F(x, \lambda)(x + u_1). \end{aligned}$$

Subtract and write

$$\begin{aligned} |Q(u_2) - Q(u_1)| &\leq |F(x + u_2, \lambda) - F(x + u_1, \lambda) - D_f F(x + u_1, \lambda)(u_2 - u_1)| \\ &\quad + |(D_f F(x + u_1, \lambda) - D_f F(x, \lambda))(u_2 - u_1)| \end{aligned}$$

for every point $x \in (0, 1]$. Thus, we only have to prove two bounds:

$$\begin{aligned} &\|F(x + u_2, \lambda) - F(x + u_1, \lambda) - D_f F(x + u_1, \lambda)(u_2 - u_1)\|_{L^\infty[0,1]} \\ &= o(1)\|u_2 - u_1\|_{\dot{L}ip[0,1]} \end{aligned} \tag{60}$$

and

$$\|D_f F(x + u, \lambda) - D_f F(x, \lambda)\|_{\dot{L}ip[0,1], L^\infty[0,1]} = o(1), \quad \|u\|_{\dot{L}ip[0,1]} \leq \delta, \quad \delta \rightarrow 0. \tag{61}$$

7.1. *The proof of (60).* We start with proving (60).

Denote $\rho(x) = x + u_1(x)$. By our assumptions we have

$$\|\rho'(x) - 1\|_{L^\infty[0,1]} \leq \delta \ll 1, \quad \rho(0) = 0.$$

Therefore,

$$\rho(x) = x \left(1 + \int_0^1 (\rho'(xt) - 1) dt \right) = x(1 + O(\delta)).$$

Remark. We will use the following property many times in the arguments below. Given arbitrary $M > 0$, the scaled function $\rho_M(\widehat{x}) = M\rho(M^{-1}\widehat{x})$ satisfies:

$$\rho_M(0) = 0, \quad \|\rho'_M(\widehat{x}) - 1\|_{L^\infty[0,M]} \leq \delta$$

Moreover, if $\|h - g\|_{\dot{L}ip[0,1]} \leq \epsilon$, then $\|h_M - g_M\|_{\dot{L}ip[0,M]} \leq \epsilon$ after scaling.

Take $t \in \mathbb{R}$ with $|t| < t_0 = \|u_2 - u_1\|_{\dot{L}ip[0,1]}$ and $f : \|f\|_{\dot{L}ip[0,1]} \leq 1$. Consider $f_t(x) = \rho(x) + tf(x)$. We only need to show that

$$\|F(f_t, \lambda) - F(f_0, \lambda) - tD_f F(f_0, \lambda)f\|_{L^\infty[0,1]} = t o(1), \quad t \rightarrow 0 \tag{62}$$

uniformly in f and ρ .

Fix arbitrary $x \in (0, 1]$ and apply the mean-value formula to $F(f_t, \lambda) - F(f_0, \lambda)$,

$$F(f_t, \lambda) - F(f_0, \lambda) = t \frac{P_1 + \dots + P_6}{x\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)}$$

where $t_1(x) \in [0, t]$. Introducing $Y_{\lambda,t}(x) = \sqrt{\lambda^2 + f_t^2(x)}$, we get

$$\begin{aligned} P_1 &= f(\rho' + t_1 f') \int_0^1 K_1(x, \tau, Y_{\lambda,t_1}) d\tau \\ P_2 &= (\rho + t_1 f) f' \int_0^1 K_1(x, \tau, Y_{\lambda,t_1}) d\tau \\ P_3 &= 2(\rho + t_1 f)(\rho' + t_1 f') (X_1 + \dots + X_4) \end{aligned}$$

where

$$\begin{aligned} X_1 &= \int_0^1 \frac{Y_{\lambda,t_1}(x) + Y_{\lambda,t_1}(\tau)}{(x + \tau)^2 + (Y_{\lambda,t_1}(x) + Y_{\lambda,t_1}(\tau))^2} \left(\frac{f_{t_1}(x)f(x)}{Y_{\lambda,t_1}(x)} + \frac{f_{t_1}(\tau)f(\tau)}{Y_{\lambda,t_1}(\tau)} \right) d\tau \\ X_2 &= \int_0^1 \frac{Y_{\lambda,t_1}(x) + Y_{\lambda,t_1}(\tau)}{(x - \tau)^2 + (Y_{\lambda,t_1}(x) + Y_{\lambda,t_1}(\tau))^2} \left(\frac{f_{t_1}(x)f(x)}{Y_{\lambda,t_1}(x)} + \frac{f_{t_1}(\tau)f(\tau)}{Y_{\lambda,t_1}(\tau)} \right) d\tau \\ X_3 &= - \int_0^1 \frac{Y_{\lambda,t_1}(x) - Y_{\lambda,t_1}(\tau)}{(x - \tau)^2 + (Y_{\lambda,t_1}(x) - Y_{\lambda,t_1}(\tau))^2} \left(\frac{f_{t_1}(x)f(x)}{Y_{\lambda,t_1}(x)} - \frac{f_{t_1}(\tau)f(\tau)}{Y_{\lambda,t_1}(\tau)} \right) d\tau \\ X_4 &= - \int_0^1 \frac{Y_{\lambda,t_1}(x) - Y_{\lambda,t_1}(\tau)}{(x + \tau)^2 + (Y_{\lambda,t_1}(x) - Y_{\lambda,t_1}(\tau))^2} \left(\frac{f_{t_1}(x)f(x)}{Y_{\lambda,t_1}(x)} - \frac{f_{t_1}(\tau)f(\tau)}{Y_{\lambda,t_1}(\tau)} \right) d\tau \end{aligned}$$

and

$$\begin{aligned} P_4 &= -\frac{f_{t_1} f}{Y_{\lambda,t_1}} \int_0^1 Y'_{\lambda,t_1}(\tau) K_2(x, \tau, Y_{\lambda,t_1}) d\tau \\ P_5 &= -Y_{\lambda,t_1} \int_0^1 \left(\frac{f_{t_1} f}{Y_{\lambda,t_1}} \right)' K_2(x, \tau, Y_{\lambda,t_1}) d\tau \\ P_6 &= -2Y_{\lambda,t_1}(L_1 + \dots + L_4). \end{aligned}$$

Similarly, for $\{L_j\}$ we have

$$\begin{aligned}
 L_1 &= \int_0^1 Y'_{\lambda,t_1}(\tau) \frac{Y_{\lambda,t_1}(x) + Y_{\lambda,t_1}(\tau)}{(x + \tau)^2 + (Y_{\lambda,t_1}(x) + Y_{\lambda,t_1}(\tau))^2} \left(\frac{f_{t_1}(x)f(x)}{Y_{\lambda,t_1}(x)} + \frac{f_{t_1}(\tau)f(\tau)}{Y_{\lambda,t_1}(\tau)} \right) d\tau \\
 L_2 &= - \int_0^1 Y'_{\lambda,t_1}(\tau) \frac{Y_{\lambda,t_1}(x) + Y_{\lambda,t_1}(\tau)}{(x - \tau)^2 + (Y_{\lambda,t_1}(x) + Y_{\lambda,t_1}(\tau))^2} \left(\frac{f_{t_1}(x)f(x)}{Y_{\lambda,t_1}(x)} + \frac{f_{t_1}(\tau)f(\tau)}{Y_{\lambda,t_1}(\tau)} \right) d\tau \\
 L_3 &= - \int_0^1 Y'_{\lambda,t_1}(\tau) \frac{Y_{\lambda,t_1}(x) - Y_{\lambda,t_1}(\tau)}{(x - \tau)^2 + (Y_{\lambda,t_1}(x) - Y_{\lambda,t_1}(\tau))^2} \left(\frac{f_{t_1}(x)f(x)}{Y_{\lambda,t_1}(x)} - \frac{f_{t_1}(\tau)f(\tau)}{Y_{\lambda,t_1}(\tau)} \right) d\tau \\
 L_4 &= \int_0^1 Y'_{\lambda,t_1}(\tau) \frac{Y_{\lambda,t_1}(x) - Y_{\lambda,t_1}(\tau)}{(x + \tau)^2 + (Y_{\lambda,t_1}(x) - Y_{\lambda,t_1}(\tau))^2} \left(\frac{f_{t_1}(x)f(x)}{Y_{\lambda,t_1}(x)} - \frac{f_{t_1}(\tau)f(\tau)}{Y_{\lambda,t_1}(\tau)} \right) d\tau
 \end{aligned}$$

We need to show that

$$\left\| \frac{(P_1 + \dots + P_6) - (P_1^0 + \dots + P_6^0)}{x\sqrt{\lambda^2 + x^2} \log^+(x^2 + \lambda^2)} \right\|_{L^\infty[0,1]} = o(1)$$

as $t \rightarrow 0$ uniformly in f and ρ . Here P_j^0 are the similar expressions taken with $t_1 = 0$.

(1) We start with $P_1 - P_1^0$.

$$\begin{aligned}
 &\frac{1}{x\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} \left| f(\rho' + t_1 f') \int_0^1 K_1(x, \tau, Y_{\lambda,t_1}) d\tau - f\rho' \int_0^1 K_1(x, \tau, Y_{\lambda,0}) d\tau \right| \\
 &\leq \frac{t}{\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} \int_0^1 K_1(x, \tau, Y_{\lambda,t_1}) d\tau \\
 &\quad + \frac{1}{\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} \left| \int_0^1 K_1(x, \tau, Y_{\lambda,t_1}) d\tau - \int_0^1 K_1(x, \tau, Y_{\lambda,0}) d\tau \right| \quad (63)
 \end{aligned}$$

To handle the first term, we use the Lemma 5.2. The Lemma 7.1 below takes care of the second term.

Lemma 7.1. *We have*

$$\left\| \frac{\int_0^1 K_1(x, \tau, Y_{\lambda,t_1}) d\tau - \int_0^1 K_1(x, \tau, Y_{\lambda,0}) d\tau}{\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} \right\|_{L^\infty[0,1]} = o(1), \quad t \rightarrow 0$$

uniformly in λ, f , and ρ .

Proof. By the mean-value formula we have

$$\int_0^1 K_1(x, \tau, Y_{\lambda,t_1}) d\tau - \int_0^1 K_1(x, \tau, Y_{\lambda,0}) d\tau = t_1(x) (\widehat{X}_1 + \dots + \widehat{X}_4)$$

where the expressions \widehat{X}_j are different from X_j defined above only by t_1 replaced with t_2 . The bound

$$\left\| \frac{\widehat{X}_1 + \dots + \widehat{X}_4}{\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} \right\|_{L^\infty[0,1]} \lesssim 1$$

follows from the Theorem 7.1 below. \square

(2) The term $P_2 - P_2^0$ can be handled in exactly the same way.

(3) The term $P_3 - P_3^0$ is more complicated.

Arguing similarly to P_1 , we only need to prove the following theorem.

Theorem 7.1.

$$\left\| \frac{(X_1 + \dots + X_4) - (X_1^0 + \dots + X_4^0)}{\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} \right\|_{L^\infty[0,1]} = o(1), \quad t \rightarrow 0$$

and

$$\left\| \frac{X_1^0 + \dots + X_4^0}{\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} \right\|_{L^\infty[0,1]} \lesssim 1$$

uniformly in λ, f , and ρ .

Proof. Let us introduce $x = \lambda \hat{x}$ and $\tau = \lambda \hat{\tau}$. Notice that

$$Y_{\lambda,t}(\lambda \hat{x}) = \lambda \sqrt{1 + (\lambda^{-1} f_t(\lambda \hat{x}))^2}.$$

Let us focus of $X_1 + X_3$ first. We are going to prove the following general result. Once we do that, it suffices to apply it to the scaled $X_1 + X_3$ by taking $y_1(x) = f(x)$ and $y_2(x) = f_{t_1}(x)$. \square

Lemma 7.2. *Suppose $y_1, y_2, \tilde{y}_2 \in \dot{L}ip[0, \lambda^{-1}]$ and*

$$\|y_1'\|_{L^\infty[0,1/\lambda]} \leq 1, \quad \|y_2' - \tilde{y}_2'\|_{L^\infty[0,1/\lambda]} \leq \epsilon, \quad \|y_2' - 1\|_{L^\infty[0,1/\lambda]} \ll 1.$$

If one defines

$$\begin{aligned} H = & \frac{1}{\sqrt{\hat{x}^2 + 1} \log^+(\lambda^2(\hat{x}^2 + 1))} \int_0^{1/\lambda} \left(\left(\frac{y_1(\hat{x})y_2(\hat{x})}{\sqrt{1 + y_2^2(\hat{x})}} + \frac{y_1(\hat{\tau})y_2(\hat{\tau})}{\sqrt{1 + y_2^2(\hat{\tau})}} \right) \right. \\ & \times \frac{\sqrt{1 + y_2^2(\hat{x})} + \sqrt{1 + y_2^2(\hat{\tau})}}{(\hat{x} + \hat{\tau})^2 + (\sqrt{1 + y_2^2(\hat{x})} + \sqrt{1 + y_2^2(\hat{\tau})})^2} - \left(\frac{y_1(\hat{x})y_2(\hat{x})}{\sqrt{1 + y_2^2(\hat{x})}} - \frac{y_1(\hat{\tau})y_2(\hat{\tau})}{\sqrt{1 + y_2^2(\hat{\tau})}} \right) \\ & \left. \times \frac{\sqrt{1 + y_2^2(\hat{x})} - \sqrt{1 + y_2^2(\hat{\tau})}}{(\hat{x} - \hat{\tau})^2 + (\sqrt{1 + y_2^2(\hat{x})} - \sqrt{1 + y_2^2(\hat{\tau})})^2} \right) d\hat{\tau} \end{aligned}$$

and

$$\begin{aligned} \tilde{H} = & \frac{1}{\sqrt{\hat{x}^2 + 1} \log^+(\lambda^2(\hat{x}^2 + 1))} \int_0^{1/\lambda} \left(\left(\frac{y_1(\hat{x})\tilde{y}_2(\hat{x})}{\sqrt{1 + \tilde{y}_2^2(\hat{x})}} + \frac{y_1(\hat{\tau})\tilde{y}_2(\hat{\tau})}{\sqrt{1 + \tilde{y}_2^2(\hat{\tau})}} \right) \right. \\ & \times \frac{\sqrt{1 + \tilde{y}_2^2(\hat{x})} + \sqrt{1 + \tilde{y}_2^2(\hat{\tau})}}{(\hat{x} + \hat{\tau})^2 + (\sqrt{1 + \tilde{y}_2^2(\hat{x})} + \sqrt{1 + \tilde{y}_2^2(\hat{\tau})})^2} - \left(\frac{y_1(\hat{x})\tilde{y}_2(\hat{x})}{\sqrt{1 + \tilde{y}_2^2(\hat{x})}} - \frac{y_1(\hat{\tau})\tilde{y}_2(\hat{\tau})}{\sqrt{1 + \tilde{y}_2^2(\hat{\tau})}} \right) \\ & \left. \times \frac{\sqrt{1 + \tilde{y}_2^2(\hat{x})} - \sqrt{1 + \tilde{y}_2^2(\hat{\tau})}}{(\hat{x} - \hat{\tau})^2 + (\sqrt{1 + \tilde{y}_2^2(\hat{x})} - \sqrt{1 + \tilde{y}_2^2(\hat{\tau})})^2} \right) d\hat{\tau} \end{aligned}$$

then

$$\|H - \tilde{H}\|_{L^\infty[0,1/\lambda]} = o(1), \quad \epsilon \rightarrow 0$$

and

$$\|H\|_{L^\infty[0,1/\lambda]} \lesssim 1$$

uniformly in $\lambda \in (0, 1)$, y_1 , y_2 , and \tilde{y}_2 .

Proof. We will study H in detail and, in particular, its stability in y_2 . That will give the necessary bounds. Notice first that

$$\left| \frac{y_2(\hat{x})}{\sqrt{y_2^2(\hat{x}) + 1}} - \frac{\tilde{y}_2(\hat{x})}{\sqrt{\tilde{y}_2^2(\hat{x}) + 1}} \right| \lesssim \begin{cases} \epsilon \hat{x}, & \hat{x} < 1 \\ \epsilon \hat{x}^{-2}, & \hat{x} > 1 \end{cases}. \tag{64}$$

The second term in the formula for H has the singularity of the type $(\hat{x} - \hat{\tau})^2$ in the denominator. However, this is compensated by the zero in the numerator and

$$\begin{aligned} \sup_{\hat{x}} \left| \int_{|\hat{\tau}-\hat{x}|<1} \left(\frac{y_1(\hat{x})y_2(\hat{x})}{\sqrt{1+y_2^2(\hat{x})}} - \frac{y_1(\hat{\tau})y_2(\hat{\tau})}{\sqrt{1+y_2^2(\hat{\tau})}} \right) \frac{\sqrt{1+y_2^2(\hat{x})}-\sqrt{1+y_2^2(\hat{\tau})}}{(\hat{x}-\hat{\tau})^2+(\sqrt{1+y_2^2(\hat{x})}-\sqrt{1+y_2^2(\hat{\tau})})^2} \right. \\ \left. - \left(\frac{y_1(\hat{x})\tilde{y}_2(\hat{x})}{\sqrt{1+\tilde{y}_2^2(\hat{x})}} - \frac{y_1(\hat{\tau})\tilde{y}_2(\hat{\tau})}{\sqrt{1+\tilde{y}_2^2(\hat{\tau})}} \right) \frac{\sqrt{1+\tilde{y}_2^2(\hat{x})}-\sqrt{1+\tilde{y}_2^2(\hat{\tau})}}{(\hat{x}-\hat{\tau})^2+(\sqrt{1+\tilde{y}_2^2(\hat{x})}-\sqrt{1+\tilde{y}_2^2(\hat{\tau})})^2} d\hat{\tau} \right| = o(1) \end{aligned}$$

when $\epsilon \rightarrow 0$ as follows from the Lemma 9.1 in Appendix. Indeed,

$$\begin{aligned} \left| \left(\frac{y_1(\hat{x})y_2(\hat{x})}{\sqrt{1+y_2^2(\hat{x})}} - \frac{y_1(\hat{x})\tilde{y}_2(\hat{x})}{\sqrt{1+\tilde{y}_2^2(\hat{x})}} \right)' \right| &\lesssim \begin{cases} \epsilon \hat{x}, & \hat{x} < 1 \\ \epsilon, & \hat{x} > 1 \end{cases} \\ \left| \left(\sqrt{1+y_2^2(\hat{x})} - \sqrt{1+\tilde{y}_2^2(\hat{x})} \right)' \right| &\lesssim \begin{cases} \epsilon \hat{x}, & \hat{x} < 1 \\ \epsilon, & \hat{x} > 1 \end{cases} \end{aligned}$$

and

$$\left| \left(\frac{y_1(\hat{x})y_2(\hat{x})}{\sqrt{1+y_2^2(\hat{x})}} \right)' \right| \lesssim \frac{\hat{x}}{\hat{x}+1}, \quad \left| \left(\sqrt{1+y_2^2(\hat{x})} \right)' \right| \lesssim \frac{\hat{x}}{\hat{x}+1}.$$

Notice also that, in the expression for H , the integral over every finite interval gives the bounded contribution after division by $\sqrt{\hat{x}^2 + 1} \log^+(\lambda^2(\hat{x}^2 + 1))$. We also have its stability in y_2 . Therefore, we can focus on $\hat{\tau} : |\hat{x} - \hat{\tau}| > 1$ only. We consider two cases: $\hat{x} \in (0, 1]$ and $\hat{x} \in [1, \lambda^{-1}]$.

(1) Let $\hat{x} \in (0, 1]$. Clearly, we can assume that $\hat{\tau} \gg 1$. Let

$$H = \frac{B_1 + B_2}{\sqrt{\hat{x}^2 + 1} \log^+(\lambda^2(\hat{x}^2 + 1))}$$

where

$$B_1 = \frac{y_1(\widehat{x})y_2(\widehat{x})}{\sqrt{1+y_2^2(\widehat{x})}} \int_0^{1/\lambda} \left(\frac{\sqrt{1+y_2^2(\widehat{x})} + \sqrt{1+y_2^2(\widehat{\tau})}}{(\widehat{x} + \widehat{\tau})^2 + (\sqrt{1+y_2^2(\widehat{x})} + \sqrt{1+y_2^2(\widehat{\tau})})^2} - \frac{\sqrt{1+y_2^2(\widehat{x})} - \sqrt{1+y_2^2(\widehat{\tau})}}{(\widehat{x} - \widehat{\tau})^2 + (\sqrt{1+y_2^2(\widehat{x})} - \sqrt{1+y_2^2(\widehat{\tau})})^2} \right) d\widehat{\tau}$$

and

$$B_2 = \int_0^{1/\lambda} \frac{y_1(\widehat{\tau})y_2(\widehat{\tau})}{\sqrt{1+y_2^2(\widehat{\tau})}} \left(\frac{\sqrt{1+y_2^2(\widehat{x})} + \sqrt{1+y_2^2(\widehat{\tau})}}{(\widehat{x} + \widehat{\tau})^2 + (\sqrt{1+y_2^2(\widehat{x})} + \sqrt{1+y_2^2(\widehat{\tau})})^2} + \frac{\sqrt{1+y_2^2(\widehat{x})} - \sqrt{1+y_2^2(\widehat{\tau})}}{(\widehat{x} - \widehat{\tau})^2 + (\sqrt{1+y_2^2(\widehat{x})} - \sqrt{1+y_2^2(\widehat{\tau})})^2} \right) d\widehat{\tau}.$$

We only need to handle integration over $\widehat{\tau} \in [2, 1/\lambda]$.

Consider B_2 first. The integrand has asymptotics

$$y_1(\widehat{\tau}) \left(2\sqrt{1+y_2^2(\widehat{x})}(\widehat{\tau}^2+y_2^2(\widehat{\tau}))^{-1} - 4y_2(\widehat{\tau}) \frac{\widehat{x}\widehat{\tau} + \sqrt{1+y_2^2(\widehat{\tau})}\sqrt{1+y_2^2(\widehat{x})}}{(\widehat{\tau}^2+y_2^2(\widehat{\tau}))^2} \right) (1+O(\widehat{\tau}^{-1})).$$

Thus, we immediately have a bound

$$|B_2| \lesssim \log^+ \lambda.$$

Comparing the integral with the one where y_2 is replaced by \widetilde{y}_2 gives us the necessary stability estimate

$$\left| \int_1^{1/\lambda} \frac{y_1(\widehat{\tau})}{\widehat{\tau}^2 + y_2^2(\widehat{\tau})} d\widehat{\tau} - \int_1^{1/\lambda} \frac{y_1(\widehat{\tau})}{\widehat{\tau}^2 + \widetilde{y}_2^2(\widehat{\tau})} d\widehat{\tau} \right| = O(\epsilon) \log^+ \lambda \tag{65}$$

and the same estimates are valid for other integrals involved. For the remainder $O(\widehat{\tau}^{-1})$, the corresponding function is bounded by $C\widehat{\tau}^{-2}$ and this decay is integrable giving a uniformly small number when integrated over $[T, 1/\lambda]$ with large T . For the integral over any finite interval $\widehat{\tau} \in [0, T]$, the stability easily follows. Thus, we first take T large and then send $\epsilon \rightarrow 0$. This will ensure the stability in y_2 .

For B_1 , the estimates are very similar. The estimate (64) gives the stability for the first factor

$$\frac{y_1(\widehat{x})y_2(\widehat{x})}{\sqrt{1+y_2^2(\widehat{x})}}$$

and the asymptotics of the integrand is $\frac{2y_2(\widehat{\tau})}{\widehat{\tau}^2 + y_2^2(\widehat{\tau})} + O(\widehat{\tau}^{-2})$. Thus, we can use an estimate similar to (65).

(2) Consider the case $\widehat{x} > 1$ now and assume that $|\widehat{x} - \widehat{\tau}| > 1$ in the integration. For $\widehat{\tau} > 1$ and $\widehat{x} > 1$, we can write

$$\begin{aligned} \sqrt{1 + y_2^2(\widehat{x})} &= y_2(\widehat{x})(1 + O(\widehat{x}^{-2})) \\ \sqrt{1 + y_2^2(\widehat{x})} + \sqrt{1 + y_2^2(\widehat{\tau})} &= (y_2(\widehat{x}) + y_2(\widehat{\tau}))R_1^{-1} \\ \sqrt{1 + y_2^2(\widehat{x})} - \sqrt{1 + y_2^2(\widehat{\tau})} &= (y_2(\widehat{x}) - y_2(\widehat{\tau}))R_1 \end{aligned} \tag{66}$$

and

$$R_1 = 1 + O\left(\frac{1}{\widehat{x}\widehat{\tau}}\right)$$

Let us control how the integral will change if we replace $\sqrt{1 + y_2^2(\widehat{x})}$ by $y_2(\widehat{x})$ and $\sqrt{1 + y_2^2(\widehat{x})} + \sqrt{1 + y_2^2(\widehat{\tau})}$ by $y_2(\widehat{x}) + y_2(\widehat{\tau})$. The errors produced in B_2 , for example, are at most

$$C_1 + C_2 \int_1^{\lambda^{-1}} \left(\frac{1}{\widehat{\tau}} + \frac{1}{\widehat{x}}\right) \frac{1}{|\widehat{x} - \widehat{\tau}| + 1} d\widehat{\tau} \lesssim 1 + \frac{\log \widehat{x} + \log^+ \lambda}{\widehat{x}}$$

The estimate for B_1 is the same. Now, notice that

$$\sup_{\widehat{x} > T, \lambda \in (0, 1)} \frac{\log \widehat{x} + \log^+ \lambda + \widehat{x}}{\widehat{x}^2 \log^+(\lambda^2(\widehat{x}^2 + 1))} \lesssim \frac{1}{\sqrt{T}} \rightarrow 0, \quad T \rightarrow \infty.$$

Since on every finite interval of integration $\widehat{\tau} \in [0, T]$ we have stability in y_2 , we only need to handle

$$\int_{\widehat{\tau} \in [0, \lambda^{-1}]} \left| \frac{(y_1(\widehat{x}) + y_1(\widehat{\tau}))(y_2(\widehat{x}) + y_2(\widehat{\tau}))}{(\widehat{x} + \widehat{\tau})^2 + (y_2(\widehat{x}) + y_2(\widehat{\tau}))^2} - \frac{(y_1(\widehat{x}) - y_1(\widehat{\tau}))(y_2(\widehat{x}) - y_2(\widehat{\tau}))}{(\widehat{x} - \widehat{\tau})^2 + (y_2(\widehat{x}) - y_2(\widehat{\tau}))^2} \right| d\widehat{\tau}.$$

Let us change the variable $\widehat{\tau} = \widehat{x}\alpha$ and introduce two functions:

$$f(\alpha, \widehat{x}) = \widehat{x}^{-1}y_1(\alpha\widehat{x}), \quad g(\alpha, \widehat{x}) = \widehat{x}^{-1}y_2(\alpha\widehat{x}). \tag{67}$$

As before, we have $f(0, \widehat{x}) = g(0, \widehat{x}) = 0$,

$$|\partial_\alpha f(\alpha, \widehat{x})| = |y_1'(\alpha\widehat{x})| \leq 1, \quad |f(\alpha, \widehat{x})| \leq \alpha$$

and

$$|\partial_\alpha g(\alpha, \widehat{x}) - 1| = |y_2'(\alpha\widehat{x}) - 1| \lesssim 1,$$

Moreover, if \widetilde{g} is the scaling of \widetilde{y}_2 , then

$$\|g' - \widetilde{g}'\|_{L^\infty[0, 1/\lambda]} \leq \epsilon.$$

These estimates are uniform in \widehat{x} . The integral takes the form

$$\widehat{x} \int_0^{1/\widehat{x}} \left| \frac{(f(1) + f(\alpha))(g(1) + g(\alpha))}{(1 + \alpha)^2 + (g(1) + g(\alpha))^2} - \frac{(f(1) - f(\alpha))(g(1) - g(\alpha))}{(1 - \alpha)^2 + (g(1) - g(\alpha))^2} \right| d\alpha.$$

We can rewrite

$$\frac{(f(1) - f(\alpha))(g(1) - g(\alpha))}{(1 - \alpha)^2 + (g(1) - g(\alpha))^2} = \frac{f(1) - f(\alpha)}{1 - \alpha} \cdot \frac{g(1) - g(\alpha)}{1 - \alpha} \frac{1}{1 + \left(\frac{g(1) - g(\alpha)}{1 - \alpha}\right)^2}$$

and the Lemma 9.1 proves stability for the interval $|\alpha - 1| < 1$. Then, the stability in g can be easily seen for every interval $\alpha \in [0, T]$ given fixed T as the corresponding error is $o(1)\hat{x}$ when $\epsilon \rightarrow 0$ and

$$o(1) \sup_{\hat{x} > 1} \frac{\hat{x}}{\sqrt{\hat{x}^2 + 1} \log^+(\lambda^2(\hat{x}^2 + 1))} = o(1).$$

For large α , we get the asymptotics

$$\begin{aligned} & \frac{(f(1) + f(\alpha))(g(1) + g(\alpha))}{(1 + \alpha)^2 + (g(1) + g(\alpha))^2} - \frac{(f(1) - f(\alpha))(g(1) - g(\alpha))}{(1 - \alpha)^2 + (g(1) - g(\alpha))^2} \\ &= \frac{-4f(\alpha)g(\alpha)(\alpha + g(1)g(\alpha))}{(\alpha^2 + g^2(\alpha))^2} + \frac{2(f(1)g(\alpha) + g(1)f(\alpha))}{\alpha^2 + g^2(\alpha)} + O(\alpha^{-2}). \end{aligned}$$

The error $O(\alpha^{-2})$ is integrable and the comparison of the leading terms to the analogous expressions with g replaced by \tilde{g} gives the error at most

$$o(1) \int_1^{1/x} \frac{d\alpha}{\alpha} = o(1) \log^+ x.$$

This leads to the error of the size

$$o(1) \frac{\hat{x} \log^+ x}{\sqrt{\hat{x}^2 + 1} \log^+(x^2 + \lambda^2)} = o(1), \quad \epsilon \rightarrow 0$$

uniformly in λ and $x > \lambda$. \square

Now, we need to handle the other combination: $X_2 + X_4$. The analysis here is nearly identical and is based on the following lemma.

Lemma 7.3. *Suppose $y_1, y_2, \tilde{y}_2 \in Lip[0, \lambda^{-1}]$ and*

$$\|y'_1\|_{L^\infty[0, 1/\lambda]} \leq 1, \quad \|y'_2 - \tilde{y}'_2\|_{L^\infty[0, 1/\lambda]} \leq \epsilon, \quad \|y'_2 - 1\|_{L^\infty[0, 1/\lambda]} \ll 1.$$

If one defines

$$\begin{aligned} H^{(1)} = & \frac{1}{\sqrt{\hat{x}^2 + 1} \log^+(\lambda^2(\hat{x}^2 + 1))} \int_0^{1/\lambda} \left(\left(\frac{y_1(\hat{x})y_2(\hat{x})}{\sqrt{1 + y_2^2(\hat{x})}} + \frac{y_1(\hat{\tau})y_2(\hat{\tau})}{\sqrt{1 + y_2^2(\hat{\tau})}} \right) \right. \\ & \times \frac{\sqrt{1 + y_2^2(\hat{x})} + \sqrt{1 + y_2^2(\hat{\tau})}}{(\hat{x} - \hat{\tau})^2 + (\sqrt{1 + y_2^2(\hat{x})} + \sqrt{1 + y_2^2(\hat{\tau})})^2} - \left(\frac{y_1(\hat{x})y_2(\hat{x})}{\sqrt{1 + y_2^2(\hat{x})}} - \frac{y_1(\hat{\tau})y_2(\hat{\tau})}{\sqrt{1 + y_2^2(\hat{\tau})}} \right) \\ & \left. \times \frac{\sqrt{1 + y_2^2(\hat{x})} - \sqrt{1 + y_2^2(\hat{\tau})}}{(\hat{x} + \hat{\tau})^2 + (\sqrt{1 + y_2^2(\hat{x})} - \sqrt{1 + y_2^2(\hat{\tau})})^2} \right) d\hat{\tau} \end{aligned}$$

and

$$\begin{aligned} \tilde{H}^{(1)} = & \frac{1}{\sqrt{\hat{x}^2 + 1} \log^+(\lambda^2(\hat{x}^2 + 1))} \int_0^{1/\lambda} \left(\left(\frac{y_1(\hat{x})\tilde{y}_2(\hat{x})}{\sqrt{1 + \tilde{y}_2^2(\hat{x})}} + \frac{y_1(\hat{\tau})\tilde{y}_2(\hat{\tau})}{\sqrt{1 + \tilde{y}_2^2(\hat{\tau})}} \right) \right. \\ & \times \frac{\sqrt{1 + \tilde{y}_2^2(\hat{x})} + \sqrt{1 + \tilde{y}_2^2(\hat{\tau})}}{(\hat{x} - \hat{\tau})^2 + (\sqrt{1 + \tilde{y}_2^2(\hat{x})} + \sqrt{1 + \tilde{y}_2^2(\hat{\tau})})^2} - \left(\frac{y_1(\hat{x})\tilde{y}_2(\hat{x})}{\sqrt{1 + \tilde{y}_2^2(\hat{x})}} - \frac{y_1(\hat{\tau})\tilde{y}_2(\hat{\tau})}{\sqrt{1 + \tilde{y}_2^2(\hat{\tau})}} \right) \\ & \left. \times \frac{\sqrt{1 + \tilde{y}_2^2(\hat{x})} - \sqrt{1 + \tilde{y}_2^2(\hat{\tau})}}{(\hat{x} + \hat{\tau})^2 + (\sqrt{1 + \tilde{y}_2^2(\hat{x})} - \sqrt{1 + \tilde{y}_2^2(\hat{\tau})})^2} \right) d\hat{\tau} \end{aligned}$$

then, uniformly in y_1, y_2, \tilde{y}_2 and $\lambda \in (0, 1)$, we have

$$\|H^{(1)} - \tilde{H}^{(1)}\|_{L^\infty[0,1/\lambda]} = o(1), \quad \epsilon \rightarrow 0$$

and

$$\|H^{(1)}\|_{L^\infty[0,1/\lambda]} \lesssim 1.$$

Proof. The proof of this lemma repeats the argument for the previous one word for word. The only minor change is contained in how we handle the singularity in the denominator of X_4 when both x and τ go to zero. After the rescaling, we have an integral

$$\begin{aligned} & \left| \int_0^1 \left(\frac{y_1(\hat{x})y_2(\hat{x})}{\sqrt{1 + y_2^2(\hat{x})}} - \frac{y_1(\hat{\tau})y_2(\hat{\tau})}{\sqrt{1 + y_2^2(\hat{\tau})}} \right) \frac{\sqrt{1 + y_2^2(\hat{x})} - \sqrt{1 + y_2^2(\hat{\tau})}}{(\hat{x} + \hat{\tau})^2 + (\sqrt{1 + y_2^2(\hat{x})} - \sqrt{1 + y_2^2(\hat{\tau})})^2} d\hat{\tau} \right| \\ & \lesssim \int_0^1 \frac{|\hat{x} - \hat{\tau}|^2}{\hat{x}^2 + \hat{\tau}^2} d\hat{\tau} \lesssim 1 \end{aligned}$$

by the application of mean-value theorem. The stability of this expression in y_2 follows from the Lemma 9.1.

This finishes the proof of Theorem 7.1. \square

We continue now with the other terms: P_4, P_5 and P_6 .

(4) Consider the term $P_4 - P_4^0$.

To study the stability in t , it is more convenient to rescale by λ and consider $y_1(\hat{x}) = \lambda^{-1} f(\hat{x}\lambda)$ and $y_2(\hat{x}) = \lambda^{-1} f_t(\hat{x}\lambda)$. Then, the problem is reduced to proving the stability of

$$P_4 = \frac{1}{\hat{x}\sqrt{1 + \hat{x}^2} \log^+(\lambda^2(\hat{x}^2 + 1))} \frac{y_1(\hat{x})y_2(\hat{x})}{\sqrt{1 + y_2^2(\hat{x})}} \int_0^{1/\lambda} \frac{y_2(\hat{\tau})y_2'(\hat{\tau})}{\sqrt{1 + y_2^2(\hat{\tau})}} K_2(\hat{x}, \hat{\tau}, \sqrt{1 + y_2^2(\hat{\tau})}) d\hat{\tau}$$

in y_2 . As before, we will be taking \tilde{y}_2 with $\|y_2' - \tilde{y}_2'\|_{L^\infty[0,1/\lambda]} \leq \epsilon$ and making a comparison.

By (64) and Lemma 5.4, we have

$$\left| y_1(\hat{x}) \left(\frac{y_2(\hat{x})}{\sqrt{1+y_2^2(\hat{x})}} - \frac{\tilde{y}_2(\hat{x})}{\sqrt{1+\tilde{y}_2^2(\hat{x})}} \right) \int_0^{1/\lambda} \frac{y_2(\hat{\tau})y_2'(\hat{\tau})}{\sqrt{1+y_2^2(\hat{\tau})}} K_2(\hat{x}, \hat{\tau}, \sqrt{1+y_2^2(\hat{\tau})}) d\hat{\tau} \right| \leq \epsilon \hat{x}^3 \log(1/\lambda), \quad \hat{x} \in (0, 1)$$

and

$$\leq \epsilon \log(1/x), \quad \hat{x} > 1$$

Thus, after division, it gives an error at most ϵ .

For the next term, (64) again gives

$$\left| \frac{y_1(\hat{x})y_2(\hat{x})}{\sqrt{1+y_2^2(\hat{x})}} \int_0^{1/\lambda} \left(\frac{y_2(\hat{\tau})y_2'(\hat{\tau})}{\sqrt{1+y_2^2(\hat{\tau})}} - \frac{\tilde{y}_2(\hat{\tau})\tilde{y}_2'(\hat{\tau})}{\sqrt{1+\tilde{y}_2^2(\hat{\tau})}} \right) K_2(\hat{x}, \hat{\tau}, \sqrt{1+y_2^2(\hat{\tau})}) d\hat{\tau} \right| \lesssim \epsilon \hat{x} \int_0^1 \hat{\tau} |K_2(\hat{x}, \hat{\tau}, \sqrt{1+y_2^2(\hat{\tau})})| d\hat{\tau} + \epsilon \hat{x} \int_1^{1/\lambda} |K_2(\hat{x}, \hat{\tau}, \sqrt{1+y_2^2(\hat{\tau})})| d\hat{\tau} \lesssim \epsilon \hat{x}^2 \log(1/\lambda), \quad \hat{x} < 1$$

and

$$\lesssim \epsilon \hat{x}^2 \log(1/x), \quad \hat{x} > 1.$$

After division by $\hat{x}\sqrt{\hat{x}^2 + 1} \log^+(\lambda^2(\hat{x}^2 + 1))$, it gives an error at most $O(\epsilon)$.

For the last term

$$\frac{y_1(\hat{x})y_2(\hat{x})}{\sqrt{1+y_2^2(\hat{x})}} \int_0^{1/\lambda} \frac{y_2(\hat{\tau})y_2'(\hat{\tau})}{\sqrt{1+y_2^2(\hat{\tau})}} \left(K_2(\hat{x}, \hat{\tau}, \sqrt{1+y_2^2(\hat{\tau})}) - K_2(\hat{x}, \hat{\tau}, \sqrt{1+\tilde{y}_2^2(\tau)}) \right) d\hat{\tau} \tag{68}$$

we can apply the mean value theorem and the resulting derivative of the kernel can be handled by the Theorem 7.2 below. As the result, the expression above can be bounded by

$$\lesssim \hat{x}^2 \log(1/\lambda) o(1), \quad \hat{x} < 1$$

and

$$\lesssim \hat{x}^2 \log(1/x) o(1), \quad \hat{x} > 1.$$

Upon division by

$$\hat{x}\sqrt{\hat{x}^2 + 1} \log^+(\lambda^2(\hat{x}^2 + 1))$$

this is at most $o(1)$.

(5) The term $P_5 - P_5^0$ can be estimated similarly.

Indeed, after scaling we have the following expression

$$\sqrt{1 + y_2^2(\hat{x})} \int_0^{1/\lambda} \left(y_1' \frac{y_2}{\sqrt{1 + y_2^2}} + y_1 y_2' (1 + y_2^2)^{-1.5} \right) K_2(\hat{x}, \hat{\tau}, \sqrt{1 + y_2^2(\hat{\tau})}) d\hat{\tau}$$

and we can repeat the steps from the previous argument.

(6) We are left to handle $P_6 - P_6^0$.

This analysis is very similar to the one performed for P_3 . However, we give details for completeness.

Theorem 7.2.

$$\left\| \frac{Y_{\lambda,t}(L_1 + \dots + L_4) - Y_{\lambda,0}(L_1^0 + \dots + L_4^0)}{x\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} \right\|_{L^\infty[0,1]} = o(1), \quad t \rightarrow 0$$

uniformly in λ .

Proof. Rescale by λ and rewrite the problem for y_1 and y_2 , as before. Notice first that

$$|\sqrt{1 + y_2^2} - \sqrt{1 + \tilde{y}_2^2}| \leq \epsilon \hat{x}^2, \quad \hat{x} < 1$$

and

$$|\sqrt{1 + y_2^2} - \sqrt{1 + \tilde{y}_2^2}| \leq \epsilon \hat{x}, \quad \hat{x} > 1$$

so we only need to show that

$$\left\| \frac{(L_1 + \dots + L_4) - (L_1^0 + \dots + L_4^0)}{x \log^+(x^2 + \lambda^2)} \right\|_{L^\infty[0,1]} = o(1), \quad t \rightarrow 0$$

and

$$\left\| \frac{L_1^0 + \dots + L_4^0}{x \log^+(x^2 + \lambda^2)} \right\|_{L^\infty[0,1]} \lesssim 1. \tag{69}$$

We group $(L_1 + L_2) - (L_1^0 + L_2^0)$ and $(L_3 + L_4) - (L_3^0 + L_4^0)$ and start with the following lemma which handles $L_3 + L_4$.

Lemma 7.4. *Suppose $y_1, y_2, \tilde{y}_2 \in \dot{L}ip[0, \lambda^{-1}]$ and*

$$\|y_1'\|_{L^\infty[0,1/\lambda]} \leq 1, \quad \|y_2' - \tilde{y}_2'\|_{L^\infty[0,1/\lambda]} \leq \epsilon, \quad \|y_2' - 1\|_{L^\infty[0,1/\lambda]} \ll 1$$

If one defines

$$U = \frac{1}{\hat{x} \log^+(\lambda^2(\hat{x}^2 + 1))} \int_0^{1/\lambda} \left| \frac{y_2(\hat{\tau})y_2'(\hat{\tau})}{\sqrt{1 + y_2^2(\hat{\tau})}} \left[\left(\frac{y_1(\hat{x})y_2(\hat{x})}{\sqrt{1 + y_2^2(\hat{x})}} - \frac{y_1(\hat{\tau})y_2(\hat{\tau})}{\sqrt{1 + y_2^2(\hat{\tau})}} \right) \right. \right. \\ \left. \left. \times \left(\frac{\sqrt{1 + y_2^2(\hat{x})} - \sqrt{1 + y_2^2(\hat{\tau})}}{(\hat{x} + \hat{\tau})^2 + (\sqrt{1 + y_2^2(\hat{x})} - \sqrt{1 + y_2^2(\hat{\tau})})^2} - \frac{\sqrt{1 + y_2^2(\hat{x})} - \sqrt{1 + y_2^2(\hat{\tau})}}{(\hat{x} - \hat{\tau})^2 + (\sqrt{1 + y_2^2(\hat{x})} - \sqrt{1 + y_2^2(\hat{\tau})})^2} \right) \right] \right|$$

$$\begin{aligned}
 & - \frac{\tilde{y}_2(\hat{\tau})\tilde{y}'_2(\hat{\tau})}{\sqrt{1+\tilde{y}_2^2(\hat{\tau})}} \left[\left(\frac{y_1(\hat{x})y_2(\hat{x})}{\sqrt{1+y_2^2(\hat{x})}} - \frac{y_1(\hat{\tau})y_2(\hat{\tau})}{\sqrt{1+y_2^2(\hat{\tau})}} \right) \left(\frac{\sqrt{1+\tilde{y}_2^2(\hat{x})} - \sqrt{1+\tilde{y}_2^2(\hat{\tau})}}{(\hat{x}+\hat{\tau})^2 + (\sqrt{1+\tilde{y}_2^2(\hat{x})} - \sqrt{1+\tilde{y}_2^2(\hat{\tau})})^2} \right. \right. \\
 & \left. \left. - \frac{\sqrt{1+\tilde{y}_2^2(\hat{x})} - \sqrt{1+\tilde{y}_2^2(\hat{\tau})}}{(\hat{x} - \hat{\tau})^2 + (\sqrt{1+\tilde{y}_2^2(\hat{x})} - \sqrt{1+\tilde{y}_2^2(\hat{\tau})})^2} \right) \right] d\hat{\tau}
 \end{aligned}$$

then, uniformly in y_1, y_2, \tilde{y}_2 and $\lambda \in (0, 1)$, we have

$$\|U\|_{L^\infty[0,1/\lambda]} = o(1), \quad \epsilon \rightarrow 0.$$

Notice that in this lemma we take an absolute value inside the integration as that will make an argument more transparent.

Proof. We first prove that

$$\begin{aligned}
 & \left\| \frac{1}{\hat{x} \log^+(\lambda^2(\hat{x}^2 + 1))} \int_0^{1/\lambda} \left| \left(\frac{y_2(\hat{\tau})y'_2(\hat{\tau})}{\sqrt{1+y_2^2(\hat{\tau})}} - \frac{\tilde{y}_2(\hat{\tau})\tilde{y}'_2(\hat{\tau})}{\sqrt{1+\tilde{y}_2^2(\hat{\tau})}} \right) \right. \right. \\
 & \times \left(\frac{y_1(\hat{x})y_2(\hat{x})}{\sqrt{1+y_2^2(\hat{x})}} - \frac{y_1(\hat{\tau})y_2(\hat{\tau})}{\sqrt{1+y_2^2(\hat{\tau})}} \right) \times \left(\frac{\sqrt{1+y_2^2(\hat{x})} - \sqrt{1+y_2^2(\hat{\tau})}}{(\hat{x} + \hat{\tau})^2 + (\sqrt{1+y_2^2(\hat{x})} - \sqrt{1+y_2^2(\hat{\tau})})^2} \right. \\
 & \left. \left. - \frac{\sqrt{1+y_2^2(\hat{x})} - \sqrt{1+y_2^2(\hat{\tau})}}{(\hat{x} - \hat{\tau})^2 + (\sqrt{1+y_2^2(\hat{x})} - \sqrt{1+y_2^2(\hat{\tau})})^2} \right) \right| d\hat{\tau} \Bigg\|_{L^\infty[0,\lambda^{-1}]} = o(1) \tag{70}
 \end{aligned}$$

as $\epsilon \rightarrow 0$, uniformly in parameters. Let us observe that

$$\left| \frac{y_2(\hat{\tau})y'_2(\hat{\tau})}{\sqrt{1+y_2^2(\hat{\tau})}} - \frac{\tilde{y}_2(\hat{\tau})\tilde{y}'_2(\hat{\tau})}{\sqrt{1+\tilde{y}_2^2(\hat{\tau})}} \right| \lesssim \epsilon \hat{\tau}, \quad \hat{\tau} < 1$$

and

$$\left| \frac{y_2(\hat{\tau})y'_2(\hat{\tau})}{\sqrt{1+y_2^2(\hat{\tau})}} - \frac{\tilde{y}_2(\hat{\tau})\tilde{y}'_2(\hat{\tau})}{\sqrt{1+\tilde{y}_2^2(\hat{\tau})}} \right| \lesssim \epsilon, \quad \hat{\tau} > 1.$$

Therefore, to show (70) it is sufficient to use an estimate (72) proved below, and the following inequality

$$\begin{aligned}
 & \frac{1}{\hat{x} \log^+(\lambda^2(\hat{x}^2 + 1))} \int_0^{1/\lambda} \left| \frac{\hat{\tau}}{\sqrt{\hat{\tau}^2 + 1}} \right. \\
 & \times \left(\frac{y_1(\hat{x})y_2(\hat{x})}{\sqrt{1+y_2^2(\hat{x})}} - \frac{y_1(\hat{\tau})y_2(\hat{\tau})}{\sqrt{1+y_2^2(\hat{\tau})}} \right) \times \left(\frac{\sqrt{1+y_2^2(\hat{x})} - \sqrt{1+y_2^2(\hat{\tau})}}{(\hat{x} + \hat{\tau})^2 + (\sqrt{1+y_2^2(\hat{x})} - \sqrt{1+y_2^2(\hat{\tau})})^2} \right. \\
 & \left. \left. - \frac{\sqrt{1+y_2^2(\hat{x})} - \sqrt{1+y_2^2(\hat{\tau})}}{(\hat{x} - \hat{\tau})^2 + (\sqrt{1+y_2^2(\hat{x})} - \sqrt{1+y_2^2(\hat{\tau})})^2} \right) \right| d\hat{\tau} \lesssim 1. \tag{71}
 \end{aligned}$$

The latter can be achieved in a standard way by following, e.g, the estimates in the proof of (72).

Now, consider

$$U^{(1)} = \frac{1}{\widehat{x} \log^+(\lambda^2(\widehat{x}^2 + 1))} \int_0^{1/\lambda} \frac{\widehat{\tau}}{\sqrt{1 + \widehat{\tau}^2}} |F(\widehat{x}, \widehat{\tau}) - F_0(\widehat{x}, \widehat{\tau})| d\widehat{\tau}$$

where

$$F = \left(\frac{y_1(\widehat{x})y_2(\widehat{x})}{\sqrt{1 + y_2^2(\widehat{x})}} - \frac{y_1(\widehat{\tau})y_2(\widehat{\tau})}{\sqrt{1 + y_2^2(\widehat{\tau})}} \right) \left(\frac{\sqrt{1 + y_2^2(\widehat{x})} - \sqrt{1 + y_2^2(\widehat{\tau})}}{(\widehat{x} + \widehat{\tau})^2 + (\sqrt{1 + y_2^2(\widehat{x})} - \sqrt{1 + y_2^2(\widehat{\tau})})^2} - \frac{\sqrt{1 + y_2^2(\widehat{x})} - \sqrt{1 + y_2^2(\widehat{\tau})}}{(\widehat{x} - \widehat{\tau})^2 + (\sqrt{1 + y_2^2(\widehat{x})} - \sqrt{1 + y_2^2(\widehat{\tau})})^2} \right)$$

and

$$F_0 = \left(\frac{y_1(\widehat{x})\widetilde{y}_2(\widehat{x})}{\sqrt{1 + \widetilde{y}_2^2(\widehat{x})}} - \frac{y_1(\widehat{\tau})\widetilde{y}_2(\widehat{\tau})}{\sqrt{1 + \widetilde{y}_2^2(\widehat{\tau})}} \right) \left(\frac{\sqrt{1 + \widetilde{y}_2^2(\widehat{x})} - \sqrt{1 + \widetilde{y}_2^2(\widehat{\tau})}}{(\widehat{x} + \widehat{\tau})^2 + (\sqrt{1 + \widetilde{y}_2^2(\widehat{x})} - \sqrt{1 + \widetilde{y}_2^2(\widehat{\tau})})^2} - \frac{\sqrt{1 + \widetilde{y}_2^2(\widehat{x})} - \sqrt{1 + \widetilde{y}_2^2(\widehat{\tau})}}{(\widehat{x} - \widehat{\tau})^2 + (\sqrt{1 + \widetilde{y}_2^2(\widehat{x})} - \sqrt{1 + \widetilde{y}_2^2(\widehat{\tau})})^2} \right).$$

We are going to prove that

$$\|U^{(1)}\|_{L^\infty[0,1/\lambda]} = o(1), \quad \epsilon \rightarrow 0. \tag{72}$$

Consider the case $\widehat{x} \in [0, 1]$. The regime $\widehat{x} \rightarrow 0$ is what makes the difference when compared to the same analysis for P_3 . Take F and rewrite it as follows

$$F = -4\widehat{x}\widehat{\tau} \left(\frac{y_1(\widehat{x})y_2(\widehat{x})}{\sqrt{1 + y_2^2(\widehat{x})}} - \frac{y_1(\widehat{\tau})y_2(\widehat{\tau})}{\sqrt{1 + y_2^2(\widehat{\tau})}} \right) \times \frac{\sqrt{1 + y_2^2(\widehat{x})} - \sqrt{1 + y_2^2(\widehat{\tau})}}{(\widehat{x} + \widehat{\tau})^2 + (\sqrt{1 + y_2^2(\widehat{x})} - \sqrt{1 + y_2^2(\widehat{\tau})})^2} \times \frac{1}{(\widehat{x} - \widehat{\tau})^2 + (\sqrt{1 + y_2^2(\widehat{x})} - \sqrt{1 + y_2^2(\widehat{\tau})})^2}.$$

The Lemma 9.1 yields

$$\int_\sigma^T |F - F_0| d\widehat{\tau} = \widehat{x}o(1), \quad \epsilon \rightarrow 0$$

for every fixed $T > \sigma > 0$. For the integration over $[0, \sigma]$, we get

$$\int_0^\sigma (|F| + |F_0|) d\widehat{\tau} \lesssim \widehat{x} \int_0^\sigma \frac{(\widehat{x} + \widehat{\tau})\widehat{\tau}}{\widehat{x}^2 + \widehat{\tau}^2} d\widehat{\tau} \lesssim \widehat{x}\sigma.$$

This gives

$$\int_0^T |F - F_0| d\widehat{\tau} = \widehat{x}o(1), \quad \epsilon \rightarrow 0.$$

Now, for $\widehat{x} \in [0, 1]$, the asymptotics for large $\widehat{\tau}$ are

$$F = -\frac{4\widehat{x}\widehat{\tau}y_1(\widehat{\tau})y_2(\widehat{\tau})}{(\widehat{\tau}^2 + y_2^2(\widehat{\tau}))^2} + O(\widehat{\tau}^{-2}), \quad F_0 = -\frac{4\widehat{x}\widehat{\tau}y_1(\widehat{\tau})\widetilde{y}_2(\widehat{\tau})}{(\widehat{\tau}^2 + \widetilde{y}_2^2(\widehat{\tau}))^2} + O(\widehat{\tau}^{-2})$$

and therefore

$$\sup_{\widehat{x} \in [0, 1]} \int_T^{1/\lambda} |F - F_0| d\widehat{\tau} = o(1)\widehat{x} \log^+ \lambda + CT^{-1}.$$

That shows $U^{(1)}$ is small uniformly in λ and $\widehat{x} \in [0, 1]$ as long as $\epsilon \rightarrow 0$. Similarly, we can handle an interval $\widehat{x} \in [0, T]$ with arbitrary large fixed T . In case of $\widehat{x} > T$, we can treat the interval $|\widehat{\tau} - \widehat{x}| < 1$ using Lemma 9.1 as before. Outside this interval, we again use (66) to get [compare with (67)]

$$\begin{aligned} & \int_1^{1/\lambda} |F - F_0| d\widehat{\tau} \\ & \lesssim \widehat{x} \int_0^{1/x} u|f(1) - f(u)| \left| \frac{g(1) - g(u)}{((1+u)^2 + (g(1) - g(u))^2)((1-u)^2 + (g(1) - g(u))^2)} \right. \\ & \quad \left. - \frac{\widetilde{g}(1) - \widetilde{g}(u)}{((1+u)^2 + (\widetilde{g}(1) - \widetilde{g}(u))^2)((1-u)^2 + (\widetilde{g}(1) - \widetilde{g}(u))^2)} \right| du + \log^+ \lambda. \end{aligned}$$

Computing the asymptotics at infinity, we obtain that the last quantity is

$$o(1)\widehat{x} \log^+ x, \quad \epsilon \rightarrow 0.$$

Then,

$$\sup_{\widehat{x} > T} \frac{o(1)\widehat{x} \log^+ x + \log^+ \lambda}{\widehat{x} \log^+(\lambda^2(\widehat{x}^2 + 1))} = o(1) + T^{-1/2}$$

as long as $T < \lambda^{-1/2}$. This bound proves that $U^{(1)}$ is small. \square

The combination $L_1 + L_2$ is handled similarly. We need the following lemma for that.

Lemma 7.5. *Suppose $y_1, y_2, \widetilde{y}_2 \in Lip[0, \lambda^{-1}]$ and*

$$\|y_1'\|_{L^\infty[0, 1/\lambda]} \leq 1, \quad \|y_2' - \widetilde{y}_2'\|_{L^\infty[0, 1/\lambda]} \leq \epsilon, \quad \|y_2' - 1\|_{L^\infty[0, 1/\lambda]} \ll 1.$$

If one defines

$$\begin{aligned} V = & \frac{1}{\widehat{x} \log^+(\lambda^2(\widehat{x}^2 + 1))} \int_0^{1/\lambda} \left| \frac{y_2(\widehat{\tau})y_2'(\widehat{\tau})}{\sqrt{1 + y_2^2(\widehat{\tau})}} \left[\left(\frac{y_1(\widehat{x})y_2(\widehat{x})}{\sqrt{1 + y_2^2(\widehat{x})}} + \frac{y_1(\widehat{\tau})y_2(\widehat{\tau})}{\sqrt{1 + y_2^2(\widehat{\tau})}} \right) \right. \right. \\ & \left. \left. \times \left(\frac{\sqrt{1 + y_2^2(\widehat{x})} + \sqrt{1 + y_2^2(\widehat{\tau})}}{(\widehat{x} + \widehat{\tau})^2 + (\sqrt{1 + y_2^2(\widehat{x})} + \sqrt{1 + y_2^2(\widehat{\tau})})^2} - \frac{\sqrt{1 + y_2^2(\widehat{x})} + \sqrt{1 + y_2^2(\widehat{\tau})}}{(\widehat{x} - \widehat{\tau})^2 + (\sqrt{1 + y_2^2(\widehat{x})} + \sqrt{1 + y_2^2(\widehat{\tau})})^2} \right) \right] \right| \end{aligned}$$

$$-\frac{\tilde{y}_2(\tilde{\tau})\tilde{y}'_2(\tilde{\tau})}{\sqrt{1+\tilde{y}_2^2(\tilde{\tau})}} \left[\left(\frac{y_1(\hat{x})\tilde{y}_2(\hat{x})}{\sqrt{1+\tilde{y}_2^2(\hat{x})}} + \frac{y_1(\tilde{\tau})\tilde{y}_2(\tilde{\tau})}{\sqrt{1+\tilde{y}_2^2(\tilde{\tau})}} \right) \left(\frac{\sqrt{1+\tilde{y}_2^2(\hat{x})} + \sqrt{1+\tilde{y}_2^2(\tilde{\tau})}}{(\hat{x}+\tilde{\tau})^2 + (\sqrt{1+\tilde{y}_2^2(\hat{x})} + \sqrt{1+\tilde{y}_2^2(\tilde{\tau})})^2} \right. \right. \\ \left. \left. - \frac{\sqrt{1+\tilde{y}_2^2(\hat{x})} + \sqrt{1+\tilde{y}_2^2(\tilde{\tau})}}{(\hat{x}-\tilde{\tau})^2 + (\sqrt{1+\tilde{y}_2^2(\hat{x})} + \sqrt{1+\tilde{y}_2^2(\tilde{\tau})})^2} \right) \right] d\tilde{\tau}$$

then, uniformly in y_1, y_2, \tilde{y}_2 and $\lambda \in (0, 1)$, we have

$$\|V\|_{L^\infty[0,1/\lambda]} = o(1), \quad \epsilon \rightarrow 0.$$

Proof. The proof of this lemma is nearly identical. It is actually easier as the singularities in the denominator are absent.

The bound (69) follows easily from the arguments given in the proofs of Lemmas 7.4 and 7.5. The proof of the Theorem 7.2 is now finished. \square

7.2. *The bound (61).* The estimate (61) was in fact already proved in the previous subsection. Indeed, recall (18). The derivative of F involves six terms: $I_1 + \dots + I_6$.

For instance, I_2 gives the following operator

$$\left(\frac{1}{x\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} f \int_0^1 K_1(x, \tau, \sqrt{\lambda^2 + f^2}) d\tau \right) v'$$

from $\dot{L}ip[0, 1]$ to $L^\infty[0, 1]$. Take $f = x + u$ where $\|u\|_{\dot{L}ip[0,1]} \leq \epsilon$. Then, one needs to show that

$$\sup_{\lambda \in (0,1), \|v\|_{\dot{L}ip[0,1]} \leq 1, \|u\|_{\dot{L}ip[0,1]} \leq \epsilon} \left\| \frac{1}{x\sqrt{x^2 + \lambda^2} \log^+(x^2 + \lambda^2)} \right. \\ \left. \times \left(f \int_0^1 K_1(x, \tau, \sqrt{\lambda^2 + f^2}) d\tau - x \int_0^1 K_1(x, \tau, \sqrt{\lambda^2 + x^2}) d\tau \right) v' \right\|_{L^\infty[0,1]} \\ = o(1), \quad \epsilon \rightarrow 0.$$

The proof of that, however, repeats the one for (63) where $\rho = x$. All other terms corresponding to $\{I_j\}_{j \neq 2}$ can be handled similarly and that gives (61).

8. The Proof of the Main Theorem and Regularity of Solutions

We start with proving Theorem 2.1.

Proof. We can rewrite the Eq. (13) as

$$\psi = \mathcal{O}\psi$$

and the items (a), (b), and (c) stated on the same page were all justified. In particular, we can choose sufficiently small δ and λ_0 such that for every $\lambda \in (0, \lambda_0)$ the operator \mathcal{O} has

the unique fixed point in $\mathcal{B}_\delta = \{\psi : \|\psi\|_{Lip[0,1]} \leq \delta\}$. It follows from the construction (and (58) in particular) that the solution

$$y(x, \lambda) = \sqrt{\lambda^2 + (x + \psi(x, \lambda))^2}$$

converges to $|x|$ as $\lambda \rightarrow 0$. Moreover, one immediately has $y(x, \lambda) \in Lip[-1, 1]$. Since y is positive, one can substitute it to the equation and get $y \in C^1[-1, 1]$. This regularity, however, will be significantly improved in the next theorem. \square

Remark. The self-similar behavior around the origin predicted by (12) is an immediate corollary of (58).

Let us prove now that the solution $y(x, \lambda)$ is actually infinitely smooth.

Theorem 8.1. *For every $\lambda \in (0, \lambda_0)$, we have $y(x, \lambda) \in C^\infty(-1, 1)$.*

Proof. The bound (11) implies that $K_1(x, \xi, y) > 0$ and thus $\int_{-1}^1 K(x, \xi, y)d\xi > 0$ as well. We have

$$y'(x, \lambda) = \frac{\int_{-1}^1 y'(\xi, \lambda)K(x, \xi, y)d\xi}{\int_{-1}^1 K(x, \xi, y)d\xi} \tag{73}$$

and one might want to differentiate this expression consecutively hoping to use the standard bootstrapping argument. Recall that

$$K(x, \xi, y) = \log\left((x + \xi)^2 + (y(x) + y(\xi))^2\right) - \log\left((x - \xi)^2 + (y(x) - y(\xi))^2\right)$$

and the first term presents no problem for bootstrapping as \log is smooth on $(0, \infty)$ and $(x + \xi)^2 + (y(x) + y(\xi))^2$ is strictly positive. However, the second term $\log\left((x - \xi)^2 + (y(x) - y(\xi))^2\right)$ might be problematic. We will show now how to handle it. Notice that all potentially singular integrals in (73) can be written as

$$\int_{-1}^1 g(\xi) \log((x - \xi)^2 + (y(x) - y(\xi))^2)d\xi, \tag{74}$$

where g is either equal to 1 or to $y'(\xi)$. The logarithm can be represented as

$$\log((x - \xi)^2 + (y(x) - y(\xi))^2) = 2 \log|x - \xi| + \log\left(1 + \left(\frac{y(x) - y(\xi)}{x - \xi}\right)^2\right).$$

Suppose we fix λ so small that the contraction mapping works. We take $H_\delta(x) = \log(\sqrt{\delta^2 + x^2})$ instead of $H(x) = \log x$ and denote the corresponding kernel by K_δ . Then, in a similar way, one can prove the existence of $y_\delta(x, \lambda)$ and $y_\delta(x, \lambda) \rightarrow y(x, \lambda)$, $\delta \rightarrow 0$ uniformly over $[-1, 1]$. Since $H_\delta \in C^\infty(-1, 1)$, we immediately get $y_\delta(x, \lambda) \in C^\infty(-1, 1)$ so the lemmas from the Appendix are applicable. We want to obtain estimates on $\|y_\delta\|_{C^n[-a,a]}$ that are uniform in δ .

To this end, proceed by induction. Our inductive assumption is that $\|y_\delta^{(n)}\|_{L^\infty[-b,b]} < C(n, b)$ with every $b : b < 1$, uniformly in δ . The contraction mapping argument gives

us this condition for $n = 1$. Now, let us show how to use the lemmas from the Appendix to cover $n = 2$. We set $\epsilon = 1/2$.

Consider

$$y'_\delta(x)P(x) = \int_{-1}^1 y'_\delta(\xi)K_\delta(x, \xi, y_\delta)d\xi \tag{75}$$

with

$$P(x) = \int_{-1}^1 K_\delta(x, \xi, y_\delta)d\xi.$$

Then,

$$\Delta_{x_1, x_2}(y'_\delta P) = (\Delta_{x_1, x_2}y'_\delta)P(x_1) + (\Delta_{x_1, x_2}P)y'_\delta(x_2)$$

and so

$$(\Delta_{x_1, x_2}y'_\delta)P(x_1) = -(\Delta_{x_1, x_2}P)y'_\delta(x_2) + \Delta_{x_1, x_2} \left(\int_{-1}^1 K_\delta(x, \xi, y_\delta)d\xi \right).$$

The first step is to show that $\|y'_\delta\|_{C^{1/2}[-b, b]}$ is bounded uniformly in δ for every $b < 1$. To this end, it is sufficient to estimate $\Delta_{x_1, x_2}y'_\delta$. Notice that P is positive and so poses no problem. The factor y'_δ is uniformly bounded by the inductive assumption. Consider

$$\Delta_{x_1, x_2}P, \quad \Delta_{x_1, x_2} \left(\int_{-1}^1 K_\delta(x, \xi, y_\delta)d\xi \right) \tag{76}$$

and focus on the terms of the form (74). In P , the function

$$\int_{-1}^1 \log|x - \xi|d\xi$$

is smooth. For

$$\int_{-1}^1 \log \left(1 + \left(\frac{y_\delta(x) - y_\delta(\xi)}{x - \xi} \right)^2 \right) d\xi$$

we apply Lemma 9.3 and an interpolation bound

$$\sup_{x_1, x_2 \in [-b, b]} \left| \frac{\Delta_{x_1, x_2}f}{|x_1 - x_2|^{1/2}} \right| \lesssim \sqrt{\|f\|_{C^1[-b, b]}\|f\|_{C[-b, b]}}$$

to get

$$\sup_{x_1, x_2 \in [-b, b]} \left| \frac{\Delta_{x_1, x_2}P}{|x_1 - x_2|^{1/2}} \right| \lesssim (\|y_\delta\|_{C^{1.5}[-b, b]})^{1/2}.$$

For the second function in (76), we argue similarly. The estimate (79) gives

$$\sup_{x_1, x_2 \in [-b, b]} \left| \frac{\Delta_{x_1, x_2} \int_{-1}^1 y'_\delta(\xi) \log|x - \xi|d\xi}{|x_1 - x_2|^{1/2}} \right| \lesssim 1$$

by the inductive assumption. Therefore, we get

$$\|y'_\delta\|_{C^{1/2}[-b,b]} \lesssim 1 + (\|y'_\delta\|_{C^{1/2}[-b,b]})^{1/2}$$

which implies the uniform bound on $\|y'_\delta\|_{C^{1/2}[-b,b]}$ for any $b < 1$.

Now, differentiate (75) to get

$$y''_\delta P + y'_\delta P' = \left(\int_{-1}^1 y'_\delta(\xi) K_\delta(x, \xi, y_\delta) d\xi \right)'.$$

We have $P' \in C[-b, b]$ by Lemma 9.3. Then,

$$\left(\int_{-1}^1 y'_\delta(\xi) K_\delta(x, \xi, y_\delta) d\xi \right)' \in C[-b, b]$$

with bounds uniform in δ as follows from Lemma 9.3 and (78). That shows $\|y_\delta\|_{C^n[-b,b]}$ is bounded uniformly in δ for $n = 2$.

For a general n , we argue similarly. Differentiation (75) $(n - 1)$ times gives

$$y_\delta^{(n)}(x) P(x) + \Omega_{n-1}(x) = \partial_x^{(n-1)} \int_{-1}^1 y'_\delta(\xi) K_\delta(x, \xi, y_\delta) d\xi.$$

Using the inductive assumption, we first show that all norms $\|y_\delta\|_{C^{n+0.5}[-b,b]}$ are bounded uniformly in δ . Then, we bootstrap that to C^{n+1} norm.

Once the δ -independent bounds for $\|y_\delta\|_{C^{(n)}[-a,a]}$ are established, we can take $\delta \rightarrow 0$. That gives $y(x, \lambda) \in C^n[-a, a]$ for every n . Indeed, there is a sequence $\{y_{\delta_j}\} \rightarrow u$ in $C^n[-a, a]$ by Arzela-Ascoli and so $u \in C^n[-a, a]$. However this includes the uniform convergence so $y = u$. Since n is arbitrary, we get the statement of the theorem. \square

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9. Appendix

Lemma 9.1. *If $\|f' - g'\|_{L^\infty[0,T]} \leq \delta$, then*

$$\left| \frac{f(x) - f(y)}{x - y} - \frac{g(x) - g(y)}{x - y} \right| \leq \delta$$

uniformly in $x, y \in [0, T]$.

Proof. Indeed, it follows from the following representation

$$\Upsilon_f(x, y) = \frac{f(x) - f(y)}{x - y} = \int_0^1 f'(y + (x - y)t) dt \tag{77}$$

\square

The next lemmas are needed to show that the solution $y(x, \lambda)$ is infinitely smooth.

Lemma 9.2. *Suppose $f \in C^\infty(-1, 1)$ and $0 < a < b \leq 1$. Then, for every $\epsilon \in (0, 1)$,*

$$\left\| \int_{-1}^1 f(\xi) \log|x - \xi| d\xi \right\|_{C^n[-a,a]} < C(n, a, b, \epsilon) (\|f\|_{C^{n-1+\epsilon}[-b,b]} + \|f\|_{L^\infty[-1,1]}) \tag{78}$$

and

$$\left\| \int_{-1}^1 f(\xi) \log|x - \xi| d\xi \right\|_{C^{n+\epsilon}[-a,a]} < C(n, a, b, \epsilon) (\|f^{(n)}\|_{L^\infty[-b,b]} + \|f\|_{L^\infty[-1,1]}) \tag{79}$$

Proof. The convolution structure of the kernel implies that it is sufficient to prove the statement for $n = 1$ only. This amounts to checking that

$$\left\| \int_{-1}^1 \frac{f(x) - f(\xi)}{x - \xi} d\xi \right\|_{C[-a,a]} \lesssim \|f\|_{C^\epsilon[-b,b]} + \|f\|_{L^\infty[-1,1]}$$

which is trivial. The estimate (79) can be obtained in a similar way. \square

Lemma 9.3. *Suppose $f(x) \in C^\infty[-1, 1]$ and $g(x) \in C[-1, 1]$. Then*

$$\left\| \int_{-1}^1 g(\xi) \log\left(1 + \left(\frac{f(x) - f(\xi)}{x - \xi}\right)^2\right) d\xi \right\|_{C^1[-1,1]} < C_\epsilon \|f\|_{C^{1+\epsilon}[-1,1]} \|g\|_{C[-1,1]}$$

with C_ϵ independent of f .

Proof. We write (77) and differentiate to get

$$\begin{aligned} & \left| \int_{-1}^1 g(\xi) \frac{2\Upsilon_f(x, \xi)}{1 + \Upsilon_f^2(x, \xi)} \left(\int_0^1 f''(\xi + (x - \xi)t) t dt \right) d\xi \right| \\ & \lesssim \|g\|_{C[-1,1]} \int_{-1}^1 \left| \int_0^1 \frac{\partial_t(f'(\xi + (x - \xi)t) - f'(\xi))}{x - \xi} t dt \right| d\xi \\ & \lesssim \|g\|_{C[-1,1]} \int_{-1}^1 \frac{\|f'\|_{C^\epsilon[-1,1]} |x - \xi|^\epsilon}{|x - \xi|} d\xi \lesssim \epsilon^{-1} \|f\|_{C^{1+\epsilon}[-1,1]} \|g\|_{C[-1,1]}. \end{aligned}$$

\square

By consecutive differentiation, one gets

Lemma 9.4. *Suppose $f(x) \in C^\infty[-1, 1]$ and $g(x) \in C[-1, 1]$. Then*

$$\begin{aligned} & \left\| \int_{-1}^1 g(\xi) \log\left(1 + \left(\frac{f(x) - f(\xi)}{x - \xi}\right)^2\right) d\xi \right\|_{C^n[-1,1]} \\ & < \left(C_n(\epsilon) \|f\|_{C^{n+\epsilon}[-1,1]} + F_n(\|f\|_{C^n[-1,1]}) \right) \|g\|_{C[-1,1]} \end{aligned}$$

where F_n is a certain function of $\|f\|_{C^n[-1,1]}$ only.

Proof. The proof is identical to the previous one. \square

Remark. The Lemmas 9.2 and 9.4 will hold true if we replace $\log x$ by $\log \sqrt{x^2 + \delta^2}$. The resulting estimates will be δ independent.

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