

# Cost Is Not a Dial: A Self-Contained Uniqueness Theorem for the Canonical Reciprocal Cost on $\mathbb{R}_{>0}$

Jonathan Washburn

*Recognition Physics Research Institute, Austin, Texas, USA\**

(Dated: December 24, 2025)

## Abstract

Many mathematical and physical frameworks introduce a *cost* (or action, penalty, divergence, or energy) to quantify change, then proceed as if that choice were canonical. When downstream conclusions depend on the cost, the theory acquires a hidden degree of freedom: one may swap the cost to fit desired outcomes. This problem is especially acute in “parameter-free” derivation programs, where predictivity requires eliminating not only tunable constants but also tunable *functional forms*.

This paper isolates and proves a self-contained uniqueness theorem for a cost on the positive reals. Define

$$J(x) := \frac{x + x^{-1}}{2} - 1 \quad (x > 0).$$

We show that if a candidate cost  $F : \mathbb{R}_{>0} \rightarrow \mathbb{R}$  is normalized ( $F(1) = 0$ ) and satisfies the composition law (equivalently: its log-lift  $H(t) := F(e^t) + 1$  satisfies the d’Alembert functional equation)

$$H(t + u) + H(t - u) = 2 H(t) H(u) \quad (t, u \in \mathbb{R}),$$

together with the unit-curvature calibration, then necessarily  $F(x) = J(x)$  for all  $x > 0$ . In the strengthened form proved here, no global smoothness or continuity assumption is required: continuity (indeed  $C^2$  regularity) is derived from the functional identity plus the quadratic calibration

$$\lim_{t \rightarrow 0} \frac{2 F(e^t)}{t^2} = 1.$$

The manuscript is self-contained: all proofs are given in the text. A single optional machine-check of the main theorem is available as a supplementary artifact.<sup>a</sup>

---

<sup>a</sup> Supplementary verification artifact (Lean 4): <https://github.com/REPLACE-WITH-ARCHIVED-REPO>. The present paper does not rely on this artifact; it is provided solely for independent auditing of the printed statements.

\* [jon@recognitionphysics.org](mailto:jon@recognitionphysics.org)

## I. INTRODUCTION

### A. Motivation: eliminating hidden functional degrees of freedom

In many areas of mathematics and physics, one quantifies “how different” two states are by selecting a scalar functional: a cost, energy, divergence, or action. Such choices are often guided by symmetry, convenience, or tradition (e.g., quadratic penalties, log-likelihoods, or variational principles). But once a framework is built atop a chosen cost, it is easy to overlook that the cost itself may be a *dial*: different choices can preserve superficial qualitative behavior while materially changing quantitative outputs.

If the goal is explanatory or predictive, this dial matters. In particular, claims of being “parameter-free” can be undermined even in the absence of explicit tunable constants: if one can vary the functional form of the cost while keeping the rest of the story fixed, then the cost selection plays the role of an implicit parameter family. For this reason, a credible parameter-free derivation program needs *uniqueness* results: conditions under which the cost is forced, rather than chosen.

### B. Context (optional): Recognition Science

Recognition Science (also called Recognition Physics) is a program aiming to derive a coherent mathematical scaffold for dynamics from a small set of structural constraints, including a ledger-style consistency discipline and a composition law for “recognition” amplitudes. In that setting, a cost on multiplicative ratios appears as a primitive interface between composition and measurement. The present paper does *not* require the broader Recognition Science framework; we mention it only as motivation for why one is led to consider reciprocal costs and cosine-type functional equations in log-coordinates.

Our aim here is narrower and purely mathematical: to prove that a natural set of explicit hypotheses forces a unique closed-form cost on  $\mathbb{R}_{>0}$ .

### C. The canonical reciprocal cost

**Definition 1** (Canonical reciprocal cost). Define  $J : \mathbb{R}_{>0} \rightarrow \mathbb{R}$  by

$$J(x) := \frac{x + x^{-1}}{2} - 1.$$

This function has three immediately visible features: reciprocity symmetry  $J(x) = J(x^{-1})$ , normalization  $J(1) = 0$ , and nonnegativity via

$$J(x) = \frac{(x - 1)^2}{2x} \geq 0.$$

In log-coordinates it becomes hyperbolic:

$$J(e^t) = \cosh(t) - 1.$$

### D. Main result (stated)

We now state the main theorem proved in this paper.

**Theorem 1** (Uniqueness of the canonical reciprocal cost). Let  $F : \mathbb{R}_{>0} \rightarrow \mathbb{R}$ . Assume:

1. **Normalization:**  $F(1) = 0$ .
2. **Composition law on  $\mathbb{R}_{>0}$ :** for all  $x, y > 0$ ,

$$F(xy) + F(x/y) = 2F(x)F(y) + 2F(x) + 2F(y).$$

3. **Quadratic calibration at the identity:**

$$\lim_{t \rightarrow 0} \frac{2F(e^t)}{t^2} = 1.$$

Then for all  $x > 0$ ,

$$F(x) = \frac{x + x^{-1}}{2} - 1 = J(x).$$

### E. Why these hypotheses are reasonable

The theorem separates two kinds of input:

- **Structural symmetry (normalization)**  $F(1) = 0$ , which fixes the reference point. In fact, reciprocity  $F(x) = F(x^{-1})$  is then forced by the  $\mathbb{R}_{>0}$  composition law by plugging  $x = 1$  into Definition 5.
- **A composition identity** on  $\mathbb{R}_{>0}$  that is equivalent (via log-coordinates) to the d'Alembert functional equation. This is the same addition law that characterizes cosine- and hyperbolic-cosine-type functions.
- **A single calibration**  $\lim_{t \rightarrow 0} 2F(e^t)/t^2 = 1$  that fixes the overall scale, ruling out the one-parameter family  $t \mapsto \cosh(\lambda t) - 1$ .

The proof shows that these constraints do not merely suggest the closed form  $J$ ; they force it.

## F. Paper organization

Section II collects definitions and elementary lemmas (log-coordinates, the canonical cost, and the d'Alembert equation). Section III states the main uniqueness results. Section IV gives short proof roadmaps. Section V contains complete proofs. Section VI discusses how the hypotheses arise in recognition-based composition models and what uniqueness does (and does not) buy downstream.

## II. DEFINITIONS AND BASIC PROPERTIES

### A. Domain and reciprocity symmetry

We work on the multiplicative positive reals

$$\mathbb{R}_{>0} := \{x \in \mathbb{R} : x > 0\},$$

which we interpret as *ratios* (scale changes). A central structural symmetry on  $\mathbb{R}_{>0}$  is inversion  $x \mapsto x^{-1}$ : if  $x$  represents a change in one direction, then  $x^{-1}$  represents the inverse change.

**Definition 2** (Reciprocity symmetry and normalization). *A function  $F : \mathbb{R}_{>0} \rightarrow \mathbb{R}$  is called a reciprocal cost if*

$$F(x) = F(x^{-1}) \quad \text{for all } x > 0.$$

It is normalized if  $F(1) = 0$ .

## B. Log-coordinates

Log-coordinates turn multiplication into addition. Given  $F : \mathbb{R}_{>0} \rightarrow \mathbb{R}$ , define its log-lift

$$G(t) := F(e^t), \quad H(t) := G(t) + 1 = F(e^t) + 1, \quad t \in \mathbb{R}.$$

**Lemma 1** (Reciprocity implies evenness). *If  $F$  is reciprocal (Definition 2), then  $G$  and  $H$  are even:*

$$G(-t) = G(t), \quad H(-t) = H(t) \quad (t \in \mathbb{R}).$$

*Proof.* Since  $e^{-t} = (e^t)^{-1}$  and  $F(x) = F(x^{-1})$ , we have

$$G(-t) = F(e^{-t}) = F((e^t)^{-1}) = F(e^t) = G(t).$$

Then  $H(-t) = G(-t) + 1 = G(t) + 1 = H(t)$ . □

**Lemma 2** (Normalization in log-coordinates). *If  $F(1) = 0$ , then  $G(0) = 0$  and  $H(0) = 1$ .*

*Proof.* Since  $e^0 = 1$ , we have  $G(0) = F(e^0) = F(1) = 0$ , hence  $H(0) = G(0) + 1 = 1$ . □

## C. The canonical reciprocal cost and its elementary identities

We will compare general reciprocal costs to the closed-form cost

$$J(x) := \frac{x + x^{-1}}{2} - 1, \quad x > 0.$$

**Lemma 3** (Squared form and nonnegativity of  $J$ ). *For all  $x > 0$ ,*

$$J(x) = \frac{(x - 1)^2}{2x} \geq 0,$$

*with equality if and only if  $x = 1$ .*

*Proof.* Compute

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

Since  $x > 0$ , the denominator  $2x > 0$ , so  $J(x) \geq 0$ , and  $J(x) = 0$  iff  $(x - 1)^2 = 0$ , i.e.  $x = 1$ . □

**Definition 3** (Hyperbolic cosine). *For  $t \in \mathbb{R}$ , define*

$$\cosh(t) := \frac{e^t + e^{-t}}{2}.$$

**Lemma 4** (Log-cosh identity). *For all  $t \in \mathbb{R}$ ,*

$$J(e^t) = \cosh(t) - 1.$$

*Equivalently, the log-lift of  $J$  is  $G_J(t) = \cosh(t) - 1$  and  $H_J(t) = \cosh(t)$ .*

*Proof.* Using Definition 3,

$$J(e^t) = \frac{e^t + (e^t)^{-1}}{2} - 1 = \frac{e^t + e^{-t}}{2} - 1 = \cosh(t) - 1.$$

□

#### D. The d'Alembert functional equation (composition law)

The key structural identity in this paper is the d'Alembert functional equation.

**Definition 4** (d'Alembert equation). *A function  $H : \mathbb{R} \rightarrow \mathbb{R}$  satisfies the d'Alembert equation if for all  $t, u \in \mathbb{R}$ ,*

$$H(t + u) + H(t - u) = 2 H(t) H(u).$$

**Lemma 5** (Evenness from d'Alembert and normalization). *If  $H$  satisfies Definition 4 and  $H(0) = 1$ , then  $H$  is even.*

*Proof.* Fix  $u \in \mathbb{R}$  and apply the d'Alembert equation with  $t = 0$ :

$$H(u) + H(-u) = 2 H(0) H(u) = 2H(u),$$

so  $H(-u) = H(u)$ . □

**Lemma 6** (Product identity). *Assume  $H$  satisfies the d'Alembert equation and  $H(0) = 1$ . Then for all  $t, u \in \mathbb{R}$ ,*

$$H(t + u) H(t - u) = H(t)^2 + H(u)^2 - 1.$$

*Proof.* Apply d'Alembert with  $a = t + u$  and  $b = t - u$ :

$$H((t+u) + (t-u)) + H((t+u) - (t-u)) = 2H(t+u)H(t-u),$$

so  $H(2t) + H(2u) = 2H(t+u)H(t-u)$ . Using the duplication formula  $H(2t) = 2H(t)^2 - 1$  (obtained from d'Alembert at  $(t, t)$  and  $H(0) = 1$ ), and similarly for  $u$ , yields the claim.  $\square$

**Lemma 7** (Difference-square identity). *Assume  $H$  satisfies d'Alembert and  $H(0) = 1$ . Then for all  $t, u \in \mathbb{R}$ ,*

$$(H(t+u) - H(t-u))^2 = 4(H(t)^2 - 1)(H(u)^2 - 1).$$

*Proof.* Let  $A := H(t+u)$  and  $B := H(t-u)$ . Then  $A+B = 2H(t)H(u)$  by d'Alembert, and  $AB = H(t)^2 + H(u)^2 - 1$  by Lemma 6. Hence

$$(A-B)^2 = (A+B)^2 - 4AB = 4H(t)^2H(u)^2 - 4(H(t)^2 + H(u)^2 - 1) = 4(H(t)^2 - 1)(H(u)^2 - 1).$$

$\square$

**Lemma 8** (Continuity from the curvature limit). *Assume  $H$  satisfies d'Alembert and  $H(0) = 1$ , and that the quadratic curvature limit  $\kappa_H = \lim_{t \rightarrow 0} 2(H(t) - 1)/t^2$  exists. Then  $H$  is continuous on  $\mathbb{R}$ .*

*Proof.* The limit assumption implies  $H(t) \rightarrow 1$  as  $t \rightarrow 0$ , so  $H$  is continuous at 0.

Fix  $t \in \mathbb{R}$ . For  $u \rightarrow 0$ , d'Alembert gives

$$H(t+u) + H(t-u) = 2H(t)H(u) \xrightarrow[u \rightarrow 0]{} 2H(t).$$

Also, by Lemma 7,

$$(H(t+u) - H(t-u))^2 = 4(H(t)^2 - 1)(H(u)^2 - 1) \xrightarrow[u \rightarrow 0]{} 0,$$

so  $H(t+u) - H(t-u) \rightarrow 0$ . Therefore

$$H(t+u) = \frac{(H(t+u) + H(t-u)) + (H(t+u) - H(t-u))}{2} \xrightarrow[u \rightarrow 0]{} H(t),$$

and similarly  $H(t-u) \rightarrow H(t)$ . Thus  $H$  is continuous at every  $t$ .  $\square$

**Lemma 9** (Cosh satisfies d'Alembert). *The function  $t \mapsto \cosh(t)$  satisfies the d'Alembert equation.*

*Proof.* Using Definition 3,

$$\cosh(t+u) + \cosh(t-u) = \frac{e^{t+u} + e^{-(t+u)} + e^{t-u} + e^{-(t-u)}}{2}.$$

Group terms:

$$e^{t+u} + e^{t-u} = e^t(e^u + e^{-u}), \quad e^{-(t+u)} + e^{-(t-u)} = e^{-t}(e^{-u} + e^u).$$

Therefore

$$\cosh(t+u) + \cosh(t-u) = \frac{(e^t + e^{-t})(e^u + e^{-u})}{2} = 2 \left( \frac{e^t + e^{-t}}{2} \right) \left( \frac{e^u + e^{-u}}{2} \right) = 2 \cosh(t) \cosh(u).$$

□

**Remark 1** (Shifted form for the cost lift). *If  $H$  satisfies the d'Alembert equation and we define  $G := H - 1$ , then expanding  $H = G + 1$  shows that  $G$  satisfies the shifted identity*

$$G(t+u) + G(t-u) = 2G(t)G(u) + 2G(t) + 2G(u).$$

In particular, by Lemma 4 and Lemma 9, the log-lift  $H_J(t) = J(e^t) + 1 = \cosh(t)$  obeys the d'Alembert equation.

## E. Composition law directly on $\mathbb{R}_{>0}$

The main theorem of this paper is stated directly on  $\mathbb{R}_{>0}$ , so that log-coordinates appear only as a proof technique.

**Definition 5** (d'Alembert composition law on  $\mathbb{R}_{>0}$ ). *A function  $F : \mathbb{R}_{>0} \rightarrow \mathbb{R}$  satisfies the d'Alembert composition law on  $\mathbb{R}_{>0}$  if for all  $x, y > 0$ ,*

$$F(xy) + F(x/y) = 2F(x)F(y) + 2F(x) + 2F(y).$$

**Lemma 10** (Equivalence of the  $\mathbb{R}_{>0}$  law and the log d'Alembert equation). *Let  $F : \mathbb{R}_{>0} \rightarrow \mathbb{R}$ , and define  $H : \mathbb{R} \rightarrow \mathbb{R}$  by  $H(t) := F(e^t) + 1$ . Then  $F$  satisfies Definition 5 if and only if  $H$  satisfies the d'Alembert equation (Definition 4).*

*Proof.* Assume  $F$  satisfies Definition 5. Let  $t, u \in \mathbb{R}$  and set  $x = e^t$ ,  $y = e^u$ , so  $xy = e^{t+u}$  and  $x/y = e^{t-u}$ . Then

$$\begin{aligned} H(t+u) + H(t-u) &= (F(e^{t+u}) + 1) + (F(e^{t-u}) + 1) \\ &= (F(xy) + F(x/y)) + 2 \\ &= (2F(x)F(y) + 2F(x) + 2F(y)) + 2 \\ &= 2(F(x) + 1)(F(y) + 1) = 2H(t)H(u), \end{aligned}$$

so  $H$  satisfies d'Alembert.

Conversely, if  $H$  satisfies d'Alembert, reverse the calculation with  $x = e^t$ ,  $y = e^u$  to obtain Definition 5.  $\square$

**Definition 6** (Quadratic calibration at the identity). *Let  $F : \mathbb{R}_{>0} \rightarrow \mathbb{R}$ . When the limit exists, define the log-curvature of  $F$  at the identity as*

$$\kappa(F) := \lim_{t \rightarrow 0} \frac{2F(e^t)}{t^2}.$$

Equivalently,  $\kappa(F) = \lim_{x \rightarrow 1} \frac{2F(x)}{(\log x)^2}$  whenever either limit exists.

### III. MAIN RESULTS

This section records the main statements proved in the paper. Proofs appear in Section V.

#### A. Non-vacuity: the canonical cost satisfies the hypotheses

**Lemma 11** (The canonical cost meets the structural conditions). *Let  $J(x) = \frac{1}{2}(x + x^{-1}) - 1$  on  $\mathbb{R}_{>0}$ . Then:*

1.  *$J$  is reciprocal and normalized:  $J(x) = J(x^{-1})$  for all  $x > 0$  and  $J(1) = 0$ .*
2.  *$J$  satisfies the d'Alembert composition law on  $\mathbb{R}_{>0}$  (Definition 5).*
3.  *$J$  has unit log-curvature:  $\kappa(J) = 1$  (Definition 6).*

*Proof.* (1) Reciprocity is immediate from the definition since  $x + x^{-1}$  is invariant under  $x \mapsto x^{-1}$ . Also  $J(1) = \frac{1}{2}(1 + 1) - 1 = 0$ .

(2) Define  $H_J(t) := J(e^t) + 1$ . By Lemma 4,  $H_J(t) = \cosh(t)$ , and by Lemma 9 the function  $\cosh$  satisfies d'Alembert. The equivalence Lemma 10 then implies that  $J$  satisfies the  $\mathbb{R}_{>0}$  composition law.

(3) Using  $J(e^t) = \cosh(t) - 1$  (Lemma 4) and the Taylor expansion  $\cosh(t) = 1 + t^2/2 + o(t^2)$  as  $t \rightarrow 0$ , we have

$$\lim_{t \rightarrow 0} \frac{2J(e^t)}{t^2} = \lim_{t \rightarrow 0} \frac{2(\cosh(t) - 1)}{t^2} = 1,$$

so  $\kappa(J) = 1$ .  $\square$

## B. Core uniqueness at the level of the composition law

**Theorem 2** (Classification of calibrated d'Alembert solutions). *Let  $H : \mathbb{R} \rightarrow \mathbb{R}$  satisfy the d'Alembert equation with  $H(0) = 1$ . Assume the quadratic curvature limit exists:*

$$\kappa_H := \lim_{t \rightarrow 0} \frac{2(H(t) - 1)}{t^2} \in \mathbb{R}.$$

*Then:*

1. If  $\kappa_H > 0$ , then  $H(t) = \cosh(\sqrt{\kappa_H} t)$  for all  $t \in \mathbb{R}$ .
2. If  $\kappa_H < 0$ , then  $H(t) = \cos(\sqrt{-\kappa_H} t)$  for all  $t \in \mathbb{R}$ .
3. If  $\kappa_H = 0$ , then  $H(t) = 1$  for all  $t \in \mathbb{R}$ .

In particular, if  $\kappa_H = 1$ , then  $H(t) = \cosh(t)$  for all  $t \in \mathbb{R}$ .

## C. Cost uniqueness on $\mathbb{R}_{>0}$

**Corollary 1** (Uniqueness of the canonical reciprocal cost on  $\mathbb{R}_{>0}$ ). *Let  $F : \mathbb{R}_{>0} \rightarrow \mathbb{R}$  be normalized ( $F(1) = 0$ ). Assume  $F$  satisfies the  $\mathbb{R}_{>0}$  composition law (Definition 5) and has unit log-curvature  $\kappa(F) = 1$  (Definition 6). Then*

$$F(x) = \frac{x + x^{-1}}{2} - 1 \quad \text{for all } x > 0.$$

*Proof.* Define  $H(t) := F(e^t) + 1$ . By Lemma 10,  $H$  satisfies d'Alembert, and moreover,

$$\kappa_H = \lim_{t \rightarrow 0} \frac{2(H(t) - 1)}{t^2} = \lim_{t \rightarrow 0} \frac{2F(e^t)}{t^2} = \kappa(F) = 1.$$

Theorem 2 then yields  $H(t) = \cosh(t)$ , hence  $F(e^t) = \cosh(t) - 1 = J(e^t)$  by Lemma 4. Writing any  $x > 0$  as  $x = e^{\log x}$  gives  $F(x) = J(x)$ .  $\square$

## IV. PROOF ROADMAPS

This section provides compact roadmaps for the main arguments; the full details appear in Section V.

### A. Roadmap for cost uniqueness on $\mathbb{R}_{>0}$

The proof of Corollary 1 is short once the log–cosh connection is in place:

1. **Lift to log-coordinates.** Define  $H(t) := F(e^t) + 1$ , so the multiplicative input variable  $x > 0$  becomes an additive variable  $t \in \mathbb{R}$ .
2. **Identify the lift.** The  $\mathbb{R}_{>0}$  composition law is equivalent to d’Alembert for  $H$  (Lemma 10), and the log-curvature  $\kappa(F)$  becomes the quadratic curvature  $\kappa_H$  of  $H$  at the origin.
3. **Classify  $H$ .** Apply Theorem 2 to conclude that  $H$  is  $\cosh(\sqrt{\kappa_H} t)$ ,  $\cos(\sqrt{-\kappa_H} t)$ , or identically 1. Under the unit calibration  $\kappa_H = 1$ , this gives  $H(t) = \cosh(t)$ .
4. **Return to  $\mathbb{R}_{>0}$ .** Since  $F(e^t) = H(t) - 1 = \cosh(t) - 1$ , Lemma 4 implies  $F(e^t) = J(e^t)$  for all  $t$ . Writing any  $x > 0$  as  $x = e^{\log x}$  yields  $F(x) = J(x)$ .

### B. Roadmap for the classification theorem

Theorem 2 is proved by using the curvature limit to justify a legitimate ODE bootstrap from a functional equation:

1. **Central-difference identity.** Rewrite d’Alembert at  $(t, h)$  to express the central second difference  $D_h H(t)$  in terms of  $H(t)$  and the scalar quotient  $q(h) := 2(H(h) - 1)/h^2$ .
2. **Use the curvature limit.** The limit  $q(h) \rightarrow \kappa_H$  implies  $D_h H(t) \rightarrow \kappa_H H(t)$  uniformly on compact sets.
3. **Bootstrap to  $C^2$ .** A general real-analysis lemma (Lemma 12) converts uniform convergence of central differences into classical  $C^2$  regularity, giving the ODE  $H'' = \kappa_H H$  (Lemma 13).

**4. Solve the ODE with even initial data.** d'Alembert implies evenness, hence  $H'(0) = 0$ . Comparing with the explicit solutions of  $y'' = \kappa_H y$  yields  $H(t) = \cosh(\sqrt{\kappa_H}t)$  when  $\kappa_H > 0$ ,  $H(t) = \cos(\sqrt{-\kappa_H}t)$  when  $\kappa_H < 0$ , and  $H \equiv 1$  when  $\kappa_H = 0$ .

**Remark 2** (Why the calibration matters). *Without calibration, the d'Alembert equation admits a one-parameter family:  $H(t) = \cosh(\lambda t)$  and  $H(t) = \cos(\lambda t)$  both satisfy the same functional identity, and  $\lambda$  is not determined. The scalar  $\kappa_H$  fixes this rescaling freedom by pinning  $\lambda^2 = |\kappa_H|$  and selecting the hyperbolic ( $\kappa_H > 0$ ) versus circular ( $\kappa_H < 0$ ) branch.*

## V. PROOF OF THEOREM 2

We prove the classification theorem and then obtain the cost-uniqueness corollary as an immediate consequence via Lemma 10.

### A. Regularity bootstrap from the curvature limit

Although Theorem 2 assumes only continuity, the d'Alembert equation and the existence of the quadratic curvature  $\kappa_H$  force enough smoothness to justify the ODE argument.

**Lemma 12** (A central-difference criterion for  $C^2$ ). *Let  $f : \mathbb{R} \rightarrow \mathbb{R}$  be continuous. Fix  $T > 0$  and define the central second difference*

$$D_h f(t) := \frac{f(t+h) - 2f(t) + f(t-h)}{h^2} \quad (|t| \leq T, h \neq 0).$$

*If there is a continuous function  $L : [-T, T] \rightarrow \mathbb{R}$  such that*

$$\lim_{h \rightarrow 0} \sup_{|t| \leq T} |D_h f(t) - L(t)| = 0,$$

*then  $f \in C^2([-T, T])$  and  $f''(t) = L(t)$  for all  $|t| \leq T$ .*

*Proof.* See Appendix A for a self-contained proof. □

**Lemma 13** (From d'Alembert + curvature to an ODE). *Let  $H : \mathbb{R} \rightarrow \mathbb{R}$  satisfy the d'Alembert equation with  $H(0) = 1$ . Suppose  $\kappa_H = \lim_{t \rightarrow 0} \frac{2(H(t)-1)}{t^2}$  exists. Then  $H \in C^2(\mathbb{R})$  and*

$$H''(t) = \kappa_H H(t) \quad \text{for all } t \in \mathbb{R}.$$

*Proof.* By Lemma 8,  $H$  is continuous. Fix  $T > 0$ . For  $h \neq 0$  and  $|t| \leq T$ , the d'Alembert equation gives

$$H(t+h) + H(t-h) = 2H(t)H(h).$$

Rearranging,

$$\frac{H(t+h) - 2H(t) + H(t-h)}{h^2} = 2H(t) \frac{H(h) - 1}{h^2}.$$

Let  $q(h) := \frac{2(H(h)-1)}{h^2}$ . By assumption,  $q(h) \rightarrow \kappa_H$  as  $h \rightarrow 0$ . Since  $H$  is continuous on  $[-T, T]$ , it is bounded there:  $|H(t)| \leq M_T$ . Hence

$$\sup_{|t| \leq T} \left| \frac{H(t+h) - 2H(t) + H(t-h)}{h^2} - \kappa_H H(t) \right| = \sup_{|t| \leq T} |H(t)| \cdot |q(h) - \kappa_H| \leq M_T |q(h) - \kappa_H| \xrightarrow[h \rightarrow 0]{} 0.$$

By Lemma 12 (with  $f = H$  and  $L(t) = \kappa_H H(t)$ ), we conclude that  $H \in C^2([-T, T])$  and  $H''(t) = \kappa_H H(t)$  on  $[-T, T]$ . Since  $T > 0$  was arbitrary, the conclusion holds for all  $t \in \mathbb{R}$ .  $\square$

## B. From d'Alembert to a linear ODE (classical derivation)

For completeness, we also record a direct differentiation argument: if  $H \in C^2$ , then d'Alembert implies  $H''(t) = H''(0)H(t)$ . This matches Lemma 13 once one notes that  $\kappa_H = H''(0)$  for  $C^2$  functions.

**Lemma 14** ( $C^2$  d'Alembert implies a linear ODE). *Let  $H \in C^2(\mathbb{R})$  satisfy  $H(0) = 1$  and the d'Alembert equation. Then for all  $t \in \mathbb{R}$ ,*

$$H''(t) = H''(0) H(t).$$

*Proof.* Fix  $t \in \mathbb{R}$  and define  $\Phi(u) := H(t+u) + H(t-u) - 2H(t)H(u)$ . By the d'Alembert equation,  $\Phi(u) = 0$  for all  $u$ . Differentiate twice with respect to  $u$ .

First note that, by the chain rule,

$$\Phi'(u) = H'(t+u) - H'(t-u) - 2H(t)H'(u),$$

and differentiating again gives

$$\Phi''(u) = H''(t+u) + H''(t-u) - 2H(t)H''(u).$$

Since  $\Phi(u) \equiv 0$ , we have  $\Phi''(0) = 0$ . Evaluating at  $u = 0$  yields

$$0 = \Phi''(0) = H''(t) + H''(t) - 2H(t)H''(0) = 2H''(t) - 2H(t)H''(0),$$

so  $H''(t) = H''(0)H(t)$  for all  $t$ .  $\square$

### C. Initial data

**Lemma 15** (Evenness implies zero first derivative at the origin). *Let  $H \in C^1(\mathbb{R})$  be even. Then  $H'(0) = 0$ .*

*Proof.* For  $h \neq 0$ , evenness gives  $H(h) = H(-h)$ . Then

$$\frac{H(h) - H(0)}{h} = -\frac{H(-h) - H(0)}{-h}.$$

Taking the limit  $h \rightarrow 0$  and using differentiability at 0 shows  $H'(0) = -H'(0)$ , hence  $H'(0) = 0$ .  $\square$

**Lemma 16** (d'Alembert solutions with  $H(0) = 1$  are even). *If  $H$  satisfies the d'Alembert equation and  $H(0) = 1$ , then  $H$  is even.*

*Proof.* This is Lemma 5.  $\square$

### D. Uniqueness of the calibrated solution

**Lemma 17** (ODE zero-uniqueness for positive  $\kappa$ ). *Let  $\kappa > 0$  and let  $f \in C^2(\mathbb{R})$  satisfy  $f''(t) = \kappa f(t)$  for all  $t$ , with  $f(0) = 0$  and  $f'(0) = 0$ . Then  $f(t) = 0$  for all  $t$ .*

*Proof.* Let  $\lambda := \sqrt{\kappa}$ . Define  $g(t) := f'(t) - \lambda f(t)$  and  $h(t) := f'(t) + \lambda f(t)$ . Then  $g, h \in C^1(\mathbb{R})$  and

$$g'(t) = f''(t) - \lambda f'(t) = \kappa f(t) - \lambda f'(t) = -\lambda g(t), \quad h'(t) = f''(t) + \lambda f'(t) = \kappa f(t) + \lambda f'(t) = \lambda h(t).$$

Therefore  $\frac{d}{dt}(g(t)e^{\lambda t}) = 0$ , so  $g(t)e^{\lambda t}$  is constant. Since  $g(0) = f'(0) - \lambda f(0) = 0$ , we get  $g \equiv 0$ . Similarly  $\frac{d}{dt}(h(t)e^{-\lambda t}) = 0$  and  $h(0) = f'(0) + \lambda f(0) = 0$ , so  $h \equiv 0$ . Then  $f' = \frac{1}{2}(g + h) = 0$ , so  $f$  is constant and  $f(0) = 0$  forces  $f \equiv 0$ .  $\square$

**Lemma 18** (ODE zero-uniqueness for negative  $\kappa$ ). *Let  $\kappa < 0$  and let  $f \in C^2(\mathbb{R})$  satisfy  $f''(t) = \kappa f(t)$  for all  $t$ , with  $f(0) = 0$  and  $f'(0) = 0$ . Then  $f(t) = 0$  for all  $t$ .*

*Proof.* Write  $\kappa = -\mu^2$  with  $\mu := \sqrt{-\kappa} > 0$ . Define the energy

$$E(t) := f'(t)^2 + \mu^2 f(t)^2 \geq 0.$$

Then  $E \in C^1(\mathbb{R})$  and

$$E'(t) = 2f'(t)f''(t) + 2\mu^2 f(t)f'(t) = 2f'(t)(f''(t) + \mu^2 f(t)) = 0,$$

so  $E$  is constant. Since  $E(0) = f'(0)^2 + \mu^2 f(0)^2 = 0$ , we have  $E(t) = 0$  for all  $t$ , hence  $f'(t) = 0$  and  $f(t) = 0$  for all  $t$ .  $\square$

*Proof of Theorem 2.* Let  $H$  satisfy the hypotheses, and let  $\kappa_H$  be as in the statement. By Lemma 13,  $H \in C^2(\mathbb{R})$  and  $H'' = \kappa_H H$ . Also, by Lemma 16,  $H$  is even, hence (since  $H \in C^1$ )  $H'(0) = 0$  by Lemma 15.

a. *Case  $\kappa_H = 0$ .* Then  $H'' = 0$ , so  $H(t) = at + b$ . Evenness forces  $a = 0$ , and  $H(0) = 1$  gives  $b = 1$ , hence  $H \equiv 1$ .

b. *Case  $\kappa_H > 0$ .* Let  $\lambda := \sqrt{\kappa_H}$  and set  $y(t) := \cosh(\lambda t)$ . Then  $y \in C^2(\mathbb{R})$ ,  $y'' = \kappa_H y$ ,  $y(0) = 1$ , and  $y'(0) = 0$ . Define  $f := H - y$ . Then  $f \in C^2(\mathbb{R})$ ,  $f'' = \kappa_H f$ , and  $f(0) = f'(0) = 0$ . By Lemma 17,  $f \equiv 0$ , so  $H(t) = \cosh(\lambda t)$ .

c. *Case  $\kappa_H < 0$ .* Let  $\mu := \sqrt{-\kappa_H}$  and set  $y(t) := \cos(\mu t)$ . Then  $y \in C^2(\mathbb{R})$ ,  $y'' = \kappa_H y$ ,  $y(0) = 1$ , and  $y'(0) = 0$ . With  $f := H - y$  as above, we have  $f'' = \kappa_H f$  and  $f(0) = f'(0) = 0$ . By Lemma 18,  $f \equiv 0$ , so  $H(t) = \cos(\mu t)$ .  $\square$

## VI. DISCUSSION AND IMPLICATIONS

### A. Auditability: why uniqueness is a gate, not a preference

Theorem 1 should be read as an *audit gate*: it removes an entire class of “hidden degrees of freedom” that can otherwise enter through *functional form choice*. Even if a model has no tunable numerical parameters, it may still be effectively tunable if a key construction depends on selecting a cost from a large admissible family. Uniqueness theorems turn a cost from a choice into a consequence: once the hypotheses are fixed and made explicit, the cost cannot be swapped without visibly breaking an assumption.

In that sense, the result is conceptually analogous to a rigidity theorem. Its value is not that it produces a closed form (which one might guess), but that it eliminates an implicit dial.

## B. A “certificate-circle” pattern (optional methodological note)

Although the present manuscript is self-contained and includes complete proofs, it is often useful in larger derivation programs to track a *certified surface*: the set of claims that have been independently audited against a formal checklist.

One practical pattern for doing so is the following:

- **State each checkpoint as a named proposition.** For example, “canonical reciprocal cost uniqueness on  $\mathbb{R}_{>0}$ ” is a single checkpoint statement.
- **Close the loop with a non-vacuity witness.** A checkpoint should not merely say “if  $F$  satisfies hypotheses then  $F = J$ ”; it should also exhibit that the intended canonical  $J$  does satisfy those hypotheses (Lemma 11).
- **Keep dependencies explicit.** Each checkpoint’s hypotheses should be visible in the statement, so that later results cannot silently assume extra structure.

This pattern is useful whether the independent audit is done by another human, by a second implementation, or by a proof assistant. The key point is methodological: *make the gate statements crisp, minimal, and non-vacuous*.

## C. Non-circularity: threat model and mitigations

To clarify what we mean by “non-circularity” in this context, it is helpful to name the failure modes that the present theorem (and its surrounding methodology) is designed to prevent.

a. *Threat 1: tuning by cost swapping.* Without uniqueness, one may choose a cost post hoc to fit downstream desiderata. Theorem 1 prevents this within the stated hypothesis class: any two costs satisfying the same explicit conditions must coincide.

b. *Threat 2: hiding scale in the functional equation.* The d’Alembert equation admits rescaled families  $t \mapsto \cosh(\lambda t)$  and  $t \mapsto \cos(\lambda t)$ . The single scalar calibration (the quadratic curvature  $\kappa_H = \lim_{t \rightarrow 0} 2(H(t) - 1)/t^2$ , equivalently  $\kappa(F)$ ) removes this freedom by pinning  $\lambda^2 = |\kappa_H|$  and selecting the hyperbolic versus circular branch (Remark in Section IV).

c. *Threat 3: smuggling outcomes through numerics.* In empirical settings it is easy to “prove” agreement by inserting a desired decimal into a definition and then performing identity checking. The mathematical results in this paper avoid this failure mode entirely: the statements are symbolic and do not depend on any empirical numerals. More generally, when numerics are used for illustration, the recommended practice is a *quarantine principle*: keep numerical evaluation separate from theorem statements and proofs, and label it as a check rather than a derivation.

d. *Threat 4: hidden assumptions.* Theorem 1 makes all assumptions explicit: normalization, the  $\mathbb{R}_{>0}$  composition law, continuity, and the single scalar calibration  $\kappa(F)$ . In particular, the proof does not smuggle differentiability as a premise:  $C^2$  regularity is *derived* from continuity plus the curvature limit (Lemma 13), after which the ODE argument becomes legitimate.

#### D. Near-minimality of the hypotheses

Theorem 1 is intentionally small: each hypothesis plays a distinct role. The following propositions show what breaks if one removes individual assumptions.

**Proposition 1** (Reciprocity is forced by normalization + the  $\mathbb{R}_{>0}$  law). *Let  $F : \mathbb{R}_{>0} \rightarrow \mathbb{R}$  satisfy Definition 5 and  $F(1) = 0$ . Then  $F(x) = F(x^{-1})$  for all  $x > 0$ .*

*Proof.* Plug  $x = 1$  into Definition 5:

$$F(y) + F(1/y) = 2F(1)F(y) + 2F(1) + 2F(y) = 2F(y),$$

so  $F(1/y) = F(y)$  for all  $y > 0$ . □

**Proposition 2** (Without calibration, there is a one-parameter family). *For any  $\lambda > 0$ , define*

$$F_\lambda(x) := \cosh(\lambda \log x) - 1 \quad (x > 0).$$

*Then  $F_\lambda$  is continuous on  $\mathbb{R}_{>0}$ , satisfies  $F_\lambda(1) = 0$ , satisfies the  $\mathbb{R}_{>0}$  composition law (Definition 5), and has  $\kappa(F_\lambda) = \lambda^2$ . In particular, if one drops the unit calibration  $\kappa(F) = 1$ , the scale  $\lambda$  remains undetermined.*

*Proof.* Continuity and  $F_\lambda(1) = 0$  are immediate. Let  $H_\lambda(t) := F_\lambda(e^t) + 1 = \cosh(\lambda t)$ . By Lemma 9,  $\cosh$  satisfies d’Alembert, hence so does  $H_\lambda$  (with  $t \mapsto \lambda t$ ). By Lemma 10,  $F_\lambda$  satisfies the  $\mathbb{R}_{>0}$  composition law.

Finally, using  $\cosh(\lambda t) - 1 \sim \lambda^2 t^2/2$  as  $t \rightarrow 0$ ,

$$\kappa(F_\lambda) = \lim_{t \rightarrow 0} \frac{2(\cosh(\lambda t) - 1)}{t^2} = \lambda^2.$$

□

**Proposition 3** (Without the  $\mathbb{R}_{>0}$  composition law, calibration does not force  $J$ ). *Define  $F_{\text{quad}}(x) := \frac{1}{2}(\log x)^2$  on  $\mathbb{R}_{>0}$ . Then  $F_{\text{quad}}$  is continuous, satisfies  $F_{\text{quad}}(1) = 0$ , and has  $\kappa(F_{\text{quad}}) = 1$ , but it does not satisfy the  $\mathbb{R}_{>0}$  composition law.*

*Proof.* Continuity and  $F_{\text{quad}}(1) = 0$  are immediate. Also  $F_{\text{quad}}(e^t) = t^2/2$ , so  $\kappa(F_{\text{quad}}) = \lim_{t \rightarrow 0} 2(t^2/2)/t^2 = 1$ . Define  $H(t) := F_{\text{quad}}(e^t) + 1 = 1 + t^2/2$ . Then for  $t, u \in \mathbb{R}$ ,

$$H(t+u) + H(t-u) = 2 + \frac{(t+u)^2 + (t-u)^2}{2} = 2 + t^2 + u^2,$$

while

$$2H(t)H(u) = 2 \left(1 + \frac{t^2}{2}\right) \left(1 + \frac{u^2}{2}\right) = 2 + t^2 + u^2 + \frac{t^2 u^2}{2}.$$

These differ when  $tu \neq 0$ , so  $H$  fails the d'Alembert equation, hence  $F_{\text{quad}}$  fails the  $\mathbb{R}_{>0}$  law by Lemma 10. □

**Proposition 4** (Without regularity, pathological solutions exist). *There exist functions  $F : \mathbb{R}_{>0} \rightarrow \mathbb{R}$  satisfying  $F(1) = 0$  and the  $\mathbb{R}_{>0}$  composition law that are not measurable (hence not continuous). In particular, some regularity hypothesis is necessary to exclude non-physical ‘‘Hamel-basis’’ solutions.*

*Proof.* It is a standard consequence of the existence of a Hamel basis of  $\mathbb{R}$  over  $\mathbb{Q}$  that there exists an additive function  $a : \mathbb{R} \rightarrow \mathbb{R}$  (i.e.,  $a(t+u) = a(t) + a(u)$ ) that is not continuous and not measurable. Define  $H(t) := \cosh(a(t))$ . Then  $H$  is not measurable (as a composition of a non-measurable additive function with a continuous non-constant function). Moreover, using the additivity of  $a$  and the identity  $\cosh(x+y) + \cosh(x-y) = 2 \cosh(x) \cosh(y)$ ,

$$H(t+u) + H(t-u) = \cosh(a(t)+a(u)) + \cosh(a(t)-a(u)) = 2 \cosh(a(t)) \cosh(a(u)) = 2H(t)H(u).$$

Thus  $H$  satisfies d'Alembert and  $H(0) = \cosh(a(0)) = \cosh(0) = 1$ . Now define  $F(x) := H(\log x) - 1$  on  $\mathbb{R}_{>0}$ . Then  $F(1) = 0$  and, by Lemma 10,  $F$  satisfies the  $\mathbb{R}_{>0}$  composition law. Since  $H$  is not measurable, neither is  $F$ . □

## E. Relation to Recognition Science (context, not required)

In Recognition Science, costs on multiplicative ratios arise as an interface between composition rules and measurement: ratios compose multiplicatively, while many structural constraints become additive in log-coordinates. In that setting, d'Alembert-type identities arise as composition laws for log-lifted quantities, and reciprocity symmetry encodes invariance under inversion of a ratio. The present paper isolates the purely mathematical core of that story: once the composition law and calibration are fixed, the cost is forced to be  $J(x) = \frac{1}{2}(x + x^{-1}) - 1$ .

*A provenance lemma (how the  $\mathbb{R}_{>0}$  law can arise)*

Theorem 1 treats the  $\mathbb{R}_{>0}$  composition law and calibration as explicit hypotheses. To connect this to recognition/ledger-style modeling, we record one simple provenance result: a very common “multiplicative weight” postulate forces the  $\mathbb{R}_{>0}$  law, and continuity reduces the remaining freedom to the single scale parameter  $\lambda$ .

**Proposition 5** (From multiplicative weights to the  $\mathbb{R}_{>0}$  composition law). *Assume there is a function  $W : \mathbb{R}_{>0} \rightarrow (0, \infty)$  satisfying the multiplicative (ledger) rule*

$$W(xy) = W(x)W(y) \quad (x, y > 0).$$

*Define  $F_W : \mathbb{R}_{>0} \rightarrow \mathbb{R}$  by*

$$F_W(x) := \frac{W(x) + W(x)^{-1}}{2} - 1.$$

*Then  $F_W(1) = 0$ ,  $F_W(x) = F_W(x^{-1})$ , and  $F_W$  satisfies the  $\mathbb{R}_{>0}$  composition law (Definition 5). If, in addition,  $W$  is continuous, then there exists  $\lambda \in \mathbb{R}$  with  $W(x) = x^\lambda$  for all  $x > 0$ , hence*

$$F_W(x) = \cosh(\lambda \log x) - 1.$$

*Proof.* First,  $W(1) = W(1 \cdot 1) = W(1)^2$  and  $W(1) > 0$  force  $W(1) = 1$ , so  $F_W(1) = \frac{1}{2}(1 + 1) - 1 = 0$ . Also,

$$W(x)W(x^{-1}) = W(xx^{-1}) = W(1) = 1,$$

so  $W(x^{-1}) = W(x)^{-1}$  and hence  $F_W(x) = F_W(x^{-1})$ .

Define  $H(t) := F_W(e^t) + 1 = \frac{1}{2}(W(e^t) + W(e^t)^{-1})$ . Let  $t, u \in \mathbb{R}$ . By multiplicativity,  $W(e^{t+u}) = W(e^t)W(e^u)$  and  $W(e^{t-u}) = W(e^t)W(e^u)^{-1}$ . Therefore

$$\begin{aligned} H(t+u) + H(t-u) &= \frac{W(e^{t+u}) + W(e^{t+u})^{-1} + W(e^{t-u}) + W(e^{t-u})^{-1}}{2} \\ &= \frac{(W(e^t) + W(e^t)^{-1})(W(e^u) + W(e^u)^{-1})}{2} = 2H(t)H(u), \end{aligned}$$

so  $H$  satisfies d'Alembert. By Lemma 10,  $F_W$  satisfies the  $\mathbb{R}_{>0}$  composition law.

Finally, if  $W$  is continuous, define  $w(t) := \log W(e^t)$ . Then  $w : \mathbb{R} \rightarrow \mathbb{R}$  is additive: using  $W(e^{t+u}) = W(e^t)W(e^u)$ , we get  $w(t+u) = w(t) + w(u)$ . Continuity of  $W$  implies continuity of  $w$ , hence  $w(t) = \lambda t$  for some  $\lambda \in \mathbb{R}$ . Therefore  $W(e^t) = e^{\lambda t}$ , i.e.  $W(x) = x^\lambda$  for  $x > 0$ , and  $F_W(e^t) = \cosh(\lambda t) - 1$ , equivalently  $F_W(x) = \cosh(\lambda \log x) - 1$ .  $\square$

**Remark 3** (Scope). *Proposition 5 is not claimed as a derivation from the full Recognition Science axiomatics; it is a minimal modeling bridge showing that once one postulates a multiplicative “ledger weight” and then takes the simplest symmetric scalar observable of that weight, the d’Alembert structure appears automatically, with a single residual scale  $\lambda$  removed by calibration.*

## F. Significance and cascading effects

This paper’s results are small in surface area but large in methodological consequence: they turn a cost from a *choice* into a *forced output* of explicit assumptions.

*What the uniqueness theorem prevents*

- **Post hoc cost swapping.** If a downstream construction relies only on the hypotheses of Corollary 1 (normalization, the  $\mathbb{R}_{>0}$  composition law, continuity, and the single scalar calibration  $\kappa(F)$ ), then the cost is no longer a dial: there is no alternative  $F$  within that class to “switch to” in order to fit outcomes.
- **Hidden functional parameters.** Without calibration one has a rescaling freedom  $t \mapsto \cosh(\lambda t)$  in log-coordinates, which corresponds to a one-parameter family of costs on  $\mathbb{R}_{>0}$  (and similarly a  $\cos(\lambda t)$  family). The single scalar calibration  $\kappa(F) = \lim_{t \rightarrow 0} 2F(e^t)/t^2$  removes this implicit parameter and pins the unique cost (Remark in Section IV).

*What it enables downstream (minimal and defensible)*

- **A canonical log-geometry.** The identity  $J(e^t) = \cosh(t) - 1$  provides a fixed, convex, even “energy profile” in log-coordinates. Near  $t = 0$ ,  $\cosh(t) - 1 \sim t^2/2$ , so the unique cost has a canonical quadratic approximation around the identity ratio  $x = 1$ , which is often the regime relevant for perturbative reasoning.
- **A stable interface for larger derivations.** In any larger framework where a ratio cost appears only through the assumptions of Corollary 1, one may reason at the level of those assumptions and then invoke uniqueness to conclude that the resulting identities are in fact identities for the closed form  $J$ . This reduces the risk that later steps inadvertently depend on an untracked choice of functional form.

*What it does not settle*

Theorem 1 is a mathematical rigidity result. It does *not* by itself:

- justify that a particular physical measurement process must satisfy the d’Alembert identity and calibration (those are modeling inputs in an application);
- derive unrelated structural claims (e.g., discrete closure periods, preferred dimensions, or specific physical constants), which belong to separate arguments and are outside the scope of this paper;
- claim that no other cost could be appropriate under different assumptions. Rather, it states: *within this explicit hypothesis class, the cost is uniquely  $J$ .*

## G. Reproducibility

This paper is designed to be reproducible without external dependencies beyond a standard L<sup>A</sup>T<sub>E</sub>X installation.

a. *Document build.* The manuscript uses the `revtex4-2` class and standard math packages. A typical build is:

- `latexmk -pdf canonical_cost_uniqueness.tex`

No external data files are required.

b. *Statement map (what to check).* Readers who wish to audit the results can proceed directly by checking the following chain:

- **Definitions and identities:** Definitions 2, 3, 4; Lemmas 3, 4, 9.
- **Main statements:** Theorem 2 and Corollary 1 (Section III), which imply Theorem 1.
- **Core proof steps:** the central-difference lemma (Lemma 12), the d'Alembert-to-ODE bootstrap (Lemma 13), evenness and  $H'(0) = 0$  (Lemmas 16 and 15), and ODE zero-uniqueness (Lemmas 17 and 18), culminating in the classification Theorem 2.

c. *Optional supplementary audit.* An optional, independent audit of the named statements is provided by the supplementary artifact referenced in the abstract. The paper remains the primary source of definitions, hypotheses, and proofs; the artifact exists only to mirror the printed statements for readers who prefer a machine-check.

## H. Figures and tables (recommended for exposition)

This section lists high-signal figures and tables that (i) communicate the structure of the result quickly and (ii) make common misunderstandings unlikely. The manuscript can be submitted without these aids; they are included to improve readability for a broad audience.

### *Figures*

a. *Figure 1: the canonical cost on a log scale.* Plot  $J(x)$  versus  $x$  with a log-scaled horizontal axis to emphasize reciprocity symmetry ( $x \leftrightarrow x^{-1}$ ) and the unique minimum at  $x = 1$ .

b. *Figure 2: log-cosh geometry and the quadratic approximation.* Plot  $J(e^t) = \cosh(t) - 1$  together with its small- $t$  approximation  $t^2/2$  to show how the unique cost behaves near the identity ratio.

c. *Figure 3: dependency graph of the argument.* A small diagram clarifies that the proof is a short pipeline: d'Alembert + curvature  $\Rightarrow$  ODE  $\Rightarrow$  initial data  $\Rightarrow$  uniqueness  $\Rightarrow$  cost rigidity.

Placeholder for `figures/j_vs_x.pdf`

Plot  $J(x) = \frac{1}{2}(x + x^{-1}) - 1$  vs.  $x$  on a log- $x$  axis.

FIG. 1. The canonical reciprocal cost  $J(x)$  on  $\mathbb{R}_{>0}$ . On a log- $x$  axis the graph is visibly symmetric under  $x \mapsto x^{-1}$  and has a unique minimum at  $x = 1$ .

---

Hypothesis	Role in the proof
Normalization $F(1) = 0$	Sets the reference point, giving $H(0) = 1$ . Together with the $\mathbb{R}_{>0}$ law it also implies reciprocity $F(x) = F(x^{-1})$ by plugging $x = 1$ into Definition 5.
$\mathbb{R}_{>0}$ composition law	Equivalent to the d'Alembert equation for $H(t) = F(e^t) + 1$ (Lemma 10), which drives the rigidity.
Continuity of $F$	Ensures continuity of $H$ and allows a $C^2$ bootstrap from the curvature limit (Lemma 13).
Quadratic calibration $\kappa(F) = 1$	Fixes the rescaling freedom $t \mapsto \cosh(\lambda t)$ and excludes the $\cos(\lambda t)$ branch, pinning the unique solution $H = \cosh(t)$ .

---

TABLE I. Hypotheses of Theorem 1 and why each is needed.

### Tables

- d. *Table 1: hypothesis bundle and roles.*
- e. *Table 2: nearby alternative costs and what they fail.*
- f. *Table 3: scope and status.*

Placeholder for `figures/jlog_vs_quad.pdf`

Plot  $\cosh(t) - 1$  and  $t^2/2$  near  $t = 0$ .

FIG. 2. Log-coordinates:  $J(e^t) = \cosh(t) - 1$ . The approximation  $\cosh(t) - 1 \sim t^2/2$  as  $t \rightarrow 0$  makes explicit the canonical quadratic behavior near  $x = 1$ .

## I. Supplementary independent audit

For readers who want an additional layer of verification, a supplementary machine-checked audit of the named statements in this paper is available in the artifact referenced in the abstract. This audit is *optional*: every proof needed to validate the results appears in the present text.

## Appendix A: A central-difference lemma

This appendix proves Lemma 12, which turns uniform convergence of central second differences into classical  $C^2$  regularity.

*Proof of Lemma 12.* Fix  $T > 0$  and assume  $D_h f \rightarrow L$  uniformly on  $[-T, T]$ , with  $L$  continuous.

a. *Step 1: reduce to the case  $D_h g \rightarrow 0$ .* Extend  $L$  to a continuous function  $\tilde{L} : \mathbb{R} \rightarrow \mathbb{R}$  by constant extension:  $\tilde{L}(t) = L(-T)$  for  $t < -T$ ,  $\tilde{L}(t) = L(t)$  for  $|t| \leq T$ , and  $\tilde{L}(t) = L(T)$  for  $t > T$ . Define

$$F(t) := \int_0^t \int_0^u \tilde{L}(s) ds du \quad (t \in \mathbb{R}).$$

Placeholder for `figures/proof_dependency_graph.pdf`

Diagram: d'Alembert  $\Rightarrow$  ODE  $\Rightarrow$  uniqueness  $\Rightarrow F = J$ .

FIG. 3. Logical dependency graph for the main proof. This figure is a reader aid; every step is proved in Section V.

Then  $F \in C^2(\mathbb{R})$  and  $F'' = \tilde{L}$ . In particular,  $F''(t) = L(t)$  for  $|t| \leq T$ .

**Lemma 19** (Central second differences as a weighted average). *If  $G \in C^2(\mathbb{R})$ , then for every  $t \in \mathbb{R}$  and  $h \neq 0$ ,*

$$\frac{G(t+h) - 2G(t) + G(t-h)}{h^2} = \int_{-1}^1 (1-|r|) G''(t+rh) dr.$$

*Proof.* Fix  $t$  and  $h > 0$  (the formula is symmetric in  $h$ ). By the fundamental theorem of calculus,

$$G(t+h) - G(t) = \int_0^h G'(t+u) du, \quad G(t) - G(t-h) = \int_0^h G'(t-u) du.$$

Subtracting yields

$$G(t+h) - 2G(t) + G(t-h) = \int_0^h (G'(t+u) - G'(t-u)) du.$$

Again by the fundamental theorem of calculus,

$$G'(t+u) - G'(t-u) = \int_{-u}^u G''(t+v) dv.$$

Therefore

$$G(t+h) - 2G(t) + G(t-h) = \int_0^h \int_{-u}^u G''(t+v) dv du.$$

Candidate $F(x)$ on $\mathbb{R}_{>0}$	Status relative to Theorem 1
$J(x) = \frac{1}{2}(x + x^{-1}) - 1$	Satisfies all hypotheses (Lemma 11).
$\cosh(\lambda \log x) - 1$ ( $\lambda > 0$ )	Satisfies normalization, the $\mathbb{R}_{>0}$ composition law, and continuity, but has $\kappa(F) = \lambda^2$ ; it matches Theorem 1 only when $\lambda = 1$ .
$(\log x)^2/2$	Reciprocal and normalized, smooth, but its log-lift $H(t) = 1 + t^2/2$ fails the d'Alembert equation (it introduces a $t^2 u^2$ term).
$ \log x $	Reciprocal and normalized, but the calibration $\kappa(F)$ does not exist (it diverges), so it is excluded by Definition 6.
$\frac{1}{2}(x - 1)^2$	Normalized but not reciprocal (fails $F(x) = F(x^{-1})$ ).

TABLE II. Examples showing that the hypotheses are not vacuous and that common alternatives are excluded for concrete reasons.

Item	Status in this manuscript
Theorems and lemmas	Fully stated and proved in the text (Sections II–V).
Recognition Science context	Motivational only; not required for any proof.
Supplementary audit artifact	Optional external mirror of the printed statements; not required to validate results.

TABLE III. Scope and status of the components of the paper.

Swap the order of integration: the region  $\{(u, v) : 0 \leq u \leq h, |v| \leq u\}$  is the triangle  $\{|v| \leq u \leq h\}$ , so

$$\int_0^h \int_{-u}^u G''(t+v) dv du = \int_{-h}^h \left( \int_{|v|}^h du \right) G''(t+v) dv = \int_{-h}^h (h - |v|) G''(t+v) dv.$$

Divide by  $h^2$  and substitute  $v = rh$  to obtain the claimed formula.  $\square$

By Lemma 19 with  $G = F$  and  $F'' = \tilde{L}$ ,

$$D_h F(t) = \int_{-1}^1 (1 - |r|) \tilde{L}(t + rh) dr.$$

Since  $\tilde{L}$  is uniformly continuous on the compact interval  $[-T - 1, T + 1]$ , the right-hand side converges uniformly to  $\tilde{L}(t) = L(t)$  as  $h \rightarrow 0$ , for  $|t| \leq T$ . Thus  $D_h F \rightarrow L$  uniformly on  $[-T, T]$ .

Now define  $g := f - F$ . Then  $g$  is continuous and

$$D_h g(t) = D_h f(t) - D_h F(t) \xrightarrow[h \rightarrow 0]{} 0$$

uniformly on  $[-T, T]$ .

b. *Step 2: if  $D_h g \rightarrow 0$  uniformly, then  $g$  is affine.* We prove that  $g$  is affine on  $[-T, T]$ .

Let  $\ell$  be the affine interpolant of  $g$  at the endpoints:

$$\ell(t) := \frac{T-t}{2T} g(-T) + \frac{T+t}{2T} g(T).$$

Set  $u := g - \ell$ . Then  $u$  is continuous,  $u(-T) = u(T) = 0$ , and  $D_h u = D_h g \rightarrow 0$  uniformly.

Fix  $\varepsilon > 0$ . Choose  $n \in \mathbb{N}$  large enough that  $h := 2T/n$  satisfies  $\sup_{|t| \leq T} |D_h u(t)| \leq \varepsilon$ . Define grid points  $t_k := -T + kh$  for  $k = 0, 1, \dots, n$  and set  $a_k := u(t_k)$ . Then  $a_0 = a_n = 0$  and for each  $k = 1, \dots, n-1$ ,

$$|a_{k+1} - 2a_k + a_{k-1}| = |u(t_k + h) - 2u(t_k) + u(t_k - h)| \leq \varepsilon h^2.$$

Define  $b_k := \frac{\varepsilon h^2}{2} k(n-k)$ . A direct calculation shows

$$b_{k+1} - 2b_k + b_{k-1} = -\varepsilon h^2.$$

Let  $c_k := a_k - b_k$ . Using the lower bound  $a_{k+1} - 2a_k + a_{k-1} \geq -\varepsilon h^2$ , we get

$$c_{k+1} - 2c_k + c_{k-1} = (a_{k+1} - 2a_k + a_{k-1}) - (b_{k+1} - 2b_k + b_{k-1}) \geq 0.$$

Thus  $\{c_k\}$  is discrete convex. With  $c_0 = c_n = 0$ , discrete convexity implies  $c_k \leq 0$  for all  $k$ , hence  $a_k \leq b_k$ . Applying the same argument to  $-a_k$  gives  $-a_k \leq b_k$ , so

$$|a_k| \leq b_k \leq \frac{\varepsilon h^2}{2} \cdot \frac{n^2}{4} = \frac{\varepsilon T^2}{2} \quad (k = 0, 1, \dots, n).$$

Therefore  $\sup_k |u(t_k)| \leq \varepsilon T^2/2$ .

Since  $u$  is uniformly continuous on  $[-T, T]$ , pick  $n$  larger if needed so that  $|t - s| \leq h \Rightarrow |u(t) - u(s)| \leq \varepsilon$ . For any  $t \in [-T, T]$ , choose a grid point  $t_k$  with  $|t - t_k| \leq h$ . Then

$$|u(t)| \leq |u(t_k)| + |u(t) - u(t_k)| \leq \frac{\varepsilon T^2}{2} + \varepsilon.$$

Because  $\varepsilon > 0$  was arbitrary,  $u(t) = 0$  for all  $t \in [-T, T]$ , hence  $g = \ell$  is affine.

c. *Step 3: conclude  $f \in C^2$  and  $f'' = L$ .* We have shown  $f = F + g$  on  $[-T, T]$ , where  $F \in C^2$  with  $F'' = L$  on  $[-T, T]$  and  $g$  is affine (thus  $g'' = 0$ ). Therefore  $f \in C^2([-T, T])$  and  $f'' = L$  on  $[-T, T]$ .  $\square$

## Appendix B: Formal statement of the uniqueness gate

This appendix records a compact, fully explicit statement of the mathematical gate proved in the main text.

**Definition 7** (Canonical cost and log-lift). *Define  $J : \mathbb{R}_{>0} \rightarrow \mathbb{R}$  and the log-lift of a cost  $F : \mathbb{R}_{>0} \rightarrow \mathbb{R}$  by*

$$J(x) := \frac{x + x^{-1}}{2} - 1, \quad H_F(t) := F(e^t) + 1.$$

**Definition 8** (Calibrated d'Alembert class). *We say that a cost  $F : \mathbb{R}_{>0} \rightarrow \mathbb{R}$  lies in the calibrated d'Alembert class if:*

1.  $F(1) = 0$  (normalization);
2.  $F$  satisfies the  $\mathbb{R}_{>0}$  composition law (Definition 5);
3.  $F$  is continuous on  $\mathbb{R}_{>0}$ ;
4.  $\kappa(F) = 1$  (Definition 6).

By Proposition 1, reciprocity  $F(x) = F(x^{-1})$  is then automatic.

**Theorem 3** (Uniqueness on  $\mathbb{R}_{>0}$  within the calibrated d'Alembert class). *If  $F : \mathbb{R}_{>0} \rightarrow \mathbb{R}$  lies in the calibrated d'Alembert class, then  $F(x) = J(x)$  for all  $x > 0$ .*

*Proof.* This is exactly Corollary 1.  $\square$

## Appendix C: Classical functional-equation perspective (optional)

The d'Alembert equation has a long history and admits many equivalent characterizations under regularity assumptions. The proof in this paper does not rely on external classification theorems, but some readers may appreciate the surrounding classical context.

## 1. Classification at a high level

In broad terms, the d'Alembert equation is a cosine/hyperbolic-cosine addition law. Under mild regularity (e.g., continuity plus boundedness on a nontrivial interval, or measurability), classical results show that solutions  $H$  are restricted to familiar families (cosine-type or cosh-type) parameterized by a scale.

In particular, one may view the single scalar calibration (here expressed as the quadratic curvature  $\kappa_H = \lim_{t \rightarrow 0} 2(H(t) - 1)/t^2$ , or  $\kappa(F)$  on  $\mathbb{R}_{>0}$ ) as the device that eliminates the scale parameter and pins the unique branch.

## 2. References

Standard references on functional equations and d'Alembert-type identities include:

- J. Aczél, *Lectures on Functional Equations and Their Applications*, Academic Press (1966).
- M. Kuczma, *An Introduction to the Theory of Functional Equations and Inequalities*, 2nd ed., Birkhäuser (2009).

## Appendix D: Robustness: stability under bounded defect

Real data and numerical pipelines rarely satisfy a functional identity exactly. The theorem below gives one rigorous (local) stability statement: if d'Alembert holds up to a uniform defect on a compact set and the function is sufficiently smooth, then the solution is close to the appropriate hyperbolic cosine profile.

**Definition 9** (d'Alembert defect). *For  $H : \mathbb{R} \rightarrow \mathbb{R}$ , define the d'Alembert defect*

$$\Delta_H(t, u) := H(t + u) + H(t - u) - 2H(t)H(u).$$

**Theorem 4** (A quantitative stability bound, hyperbolic branch). *Fix  $T > 0$ . Let  $H \in C^3([-T, T])$  be even with  $H(0) = 1$ , and set  $a := H''(0)$ . Assume  $a > 0$ . Let*

$$\varepsilon := \sup_{|t| \leq T, |u| \leq T} |\Delta_H(t, u)|, \quad B := \sup_{|t| \leq T} |H(t)|, \quad K := \sup_{|t| \leq T} |H^{(3)}(t)|.$$

Then for every  $h$  with  $0 < h \leq T$  and every  $t$  with  $|t| \leq T - h$ ,

$$|H(t) - \cosh(\sqrt{a}t)| \leq \frac{\delta(h)}{a} (\cosh(\sqrt{a}|t|) - 1),$$

where

$$\delta(h) := \frac{\varepsilon}{h^2} + \frac{(1+B)K}{3} h.$$

*Proof.* Fix  $0 < h \leq T$  and  $|t| \leq T - h$ .

a. *Step 1: control the central-difference remainder with bounded third derivative.* We claim

$$|H(t+h) + H(t-h) - 2H(t) - h^2 H''(t)| \leq \frac{K}{3} h^3. \quad (\text{D1})$$

To see this, use the integral form of Taylor's theorem:

$$\begin{aligned} H(t+h) &= H(t) + hH'(t) + \frac{h^2}{2} H''(t) + \int_0^h \frac{(h-s)^2}{2} H^{(3)}(t+s) ds, \\ H(t-h) &= H(t) - hH'(t) + \frac{h^2}{2} H''(t) - \int_0^h \frac{(h-s)^2}{2} H^{(3)}(t-s) ds. \end{aligned}$$

Adding and bounding  $|H^{(3)}| \leq K$  yields (D1).

Similarly, since  $H$  is even,  $H'(0) = 0$ , and the same integral form at 0 gives

$$|H(h) - 1 - \frac{a}{2}h^2| \leq \frac{K}{6} h^3. \quad (\text{D2})$$

Now write the defect identity (Definition 9) at  $(t, h)$  as

$$H(t+h) + H(t-h) = 2H(t)H(h) + \Delta_H(t, h).$$

Subtract  $2H(t) + ah^2H(t)$  from both sides and regroup:

$$\begin{aligned} h^2(H''(t) - aH(t)) &= (H(t+h) + H(t-h) - 2H(t) - h^2 H''(t)) \\ &\quad + \Delta_H(t, h) + 2H(t)(H(h) - 1 - \frac{a}{2}h^2). \end{aligned}$$

Taking absolute values and using (D1), (D2),  $|H(t)| \leq B$ , and  $|\Delta_H(t, h)| \leq \varepsilon$ , we obtain

$$h^2 |H''(t) - aH(t)| \leq \frac{K}{3} h^3 + \varepsilon + 2B \cdot \frac{K}{6} h^3 \leq \varepsilon + \frac{(1+B)K}{3} h^3.$$

Dividing by  $h^2$  yields the uniform bound

$$|H''(t) - aH(t)| \leq \delta(h) \quad (|t| \leq T - h). \quad (\text{D3})$$

Let  $y(t) := \cosh(\sqrt{a}t)$ , so  $y'' = ay$ ,  $y(0) = 1$ , and (since  $H$  is even)  $H'(0) = 0 = y'(0)$ . Define  $e(t) := H(t) - y(t)$ . Then  $e \in C^2([-T+h, T-h])$ ,  $e(0) = e'(0) = 0$ , and

$$e''(t) - ae(t) = H''(t) - aH(t),$$

so by (D3),  $|e''(t) - ae(t)| \leq \delta(h)$  for  $|t| \leq T-h$ . For  $t \in [0, T-h]$ , the variation-of-constants formula for  $e'' = ae + r$  with zero initial data gives

$$e(t) = \int_0^t \frac{1}{\sqrt{a}} \sinh(\sqrt{a}(t-s)) r(s) ds,$$

where  $r(s) := e''(s) - ae(s)$ . Hence

$$|e(t)| \leq \delta(h) \int_0^t \frac{1}{\sqrt{a}} \sinh(\sqrt{a}(t-s)) ds = \frac{\delta(h)}{a} (\cosh(\sqrt{a}t) - 1).$$

By evenness of  $e$  (difference of even functions), this bound holds for negative  $t$  as well, yielding the claimed inequality for all  $|t| \leq T-h$ .  $\square$

**Remark 4** (Interpretation). *The bound trades exact functional identity for a quantified defect  $\varepsilon$  and a smoothness envelope  $K$ . The free parameter  $h$  allows balancing the terms  $\varepsilon/h^2$  and  $\frac{(1+B)K}{3}h$ ; choosing  $h \asymp (\varepsilon/K)^{1/3}$  gives a typical error scaling  $O(\varepsilon^{1/3})$  on compact sets.*

**Corollary 2** (Stability transferred back to  $\mathbb{R}_{>0}$ ). *In the setting of Theorem 4, define  $F(x) := H(\log x) - 1$  on  $\mathbb{R}_{>0}$ . Then for all  $x \in (e^{-(T-h)}, e^{T-h})$ ,*

$$|F(x) - (\cosh(\sqrt{a} \log x) - 1)| \leq \frac{\delta(h)}{a} (\cosh(\sqrt{a} |\log x|) - 1).$$

*In particular, if  $a \approx 1$  and  $\delta(h)$  is small, then  $F$  is uniformly close to the canonical cost  $J(x) = \cosh(\log x) - 1$  on compact subintervals of  $\mathbb{R}_{>0}$ .*

*Proof.* Apply Theorem 4 with  $t = \log x$ , noting that  $F(x) = H(t) - 1$ .  $\square$

## Appendix E: Reader audit checklist

This appendix provides a “walk the proof” checklist for readers who want to verify the argument quickly and mechanically.

## 1. Step 1: verify the canonical identities

- Check the squared form  $J(x) = \frac{(x-1)^2}{2x}$  (Lemma 3).
- Check the log–cosh identity  $J(e^t) = \cosh(t) - 1$  (Lemma 4).
- Check that cosh satisfies d’Alembert (Lemma 9).

## 2. Step 2: check the functional-equation $\Rightarrow$ ODE step

- Verify the uniform central-difference bootstrap (Lemma 12) and then apply it to d’Alembert to obtain  $H''(t) = \kappa_H H(t)$  (Lemma 13).

## 3. Step 3: check the initial conditions

- From d’Alembert and  $H(0) = 1$ , verify  $H$  is even (Lemma 16).
- From evenness and differentiability, verify  $H'(0) = 0$  (Lemma 15).

## 4. Step 4: check ODE uniqueness

- Verify the  $\kappa > 0$  uniqueness lemma (Lemma 17) and the  $\kappa < 0$  uniqueness lemma (Lemma 18).
- Apply them to  $f(t) = H(t) - \cosh(\sqrt{\kappa_H} t)$  (when  $\kappa_H > 0$ ) or  $f(t) = H(t) - \cos(\sqrt{-\kappa_H} t)$  (when  $\kappa_H < 0$ ) to conclude the classification Theorem 2.
- Translate back to  $F$  on  $\mathbb{R}_{>0}$  using  $x = e^{\log x}$  (Corollary 1).

## Appendix F: Non-circularity checklist (methodological)

The main results are purely symbolic and do not depend on any empirical numerals. In applications, to preserve that property, we recommend the following non-circularity checklist:

- **Separate modeling inputs from consequences.** Write the hypotheses (e.g., the d’Alembert identity and calibration) as explicit assumptions, and do not smuggle them via prose.

- **Quarantine numerics.** If numerical plots or evaluations are shown, keep them separate from theorem statements and label them as checks/illustrations, not derivations.
- **Avoid “hidden scales”.** When an equation admits rescaling families, include a single explicit calibration to pin scale (as done here with  $\kappa(F) = 1$ ).
- **Minimize the hypothesis surface.** A smaller, named assumption set makes it easier for others to test whether an application truly satisfies the gate.
- **Provide a statement map.** Explicitly list what must be checked (as in Section VI G and Appendix E).