

P vs NP via the Computation/Recognition Split: A Dual-Complexity Framework from Ledger Dynamics

An Exploratory Paper in Recognition Science

SCAFFOLD — Not a claim to have resolved P vs NP

Jonathan Washburn

Recognition Science Research Institute, Austin, Texas

washburn.jonathan@gmail.com

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Abstract

Claim hygiene. This paper explores a *hypothetical* framework for understanding the P vs NP problem; it does **not** claim to resolve it unconditionally. All separation results are conditional on the ledger-computation model.

We observe that the Turing machine model implicitly assumes zero-cost observation: reading a tape cell is free. In Recognition Science, observation has fundamental cost $J(r)$ per query. This motivates a dual-complexity framework with two independent measures:

- $T_c(n)$: *computation complexity* — internal evolution steps (double-entry ledger updates).
- $T_r(n)$: *recognition complexity* — observation operations (extracting information from the ledger).

In the standard Turing model, $T_r = 0$ and total cost = T_c . In the ledger model, $T_r > 0$ and total cost = $T_c + T_r$.

Under ledger assumptions, the double-entry structure forces *balanced-parity encoding*: information is hidden in the parity balance of ledger entries. Extracting one bit of parity requires $\Omega(n)$ recognition queries (information-theoretic lower bound). Meanwhile, the internal evolution (computation) can reorganise the ledger in $O(n^{1/3} \log n)$ steps (subpolynomial in n).

This creates a conditional separation: $T_c(\text{SAT}) = O(n^{1/3} \log n)$ but $T_r(\text{SAT}) = \Omega(n)$. The P vs NP question splits: $P = NP$ at the computation scale (internal evolution is fast), $P \neq NP$ at the recognition scale (observation is expensive). The Clay Millennium problem, as traditionally stated, conflates T_c and T_r .

Status: SCAFFOLD. The Lean formalisation (`IndisputableMonolith.Complexity.*`) uses explicit hypotheses and placeholder types. No unconditional mathematical claims are made.

Keywords: P vs NP, dual complexity, computation, recognition, ledger, balanced parity, Turing incompleteness.

Contents

1 Introduction and Claim Hygiene

What this paper claims.

1. The Turing model assumes $T_r = 0$ (zero observation cost). This is a modelling choice, not a physical law.
2. If observation has cost ($T_r > 0$), a separation between computation and recognition complexity can arise.
3. The RS ledger structure provides a concrete model in which this separation is natural.

What this paper does not claim.

- × An unconditional proof that $P \neq NP$ (or $P = NP$).
- × That the ledger model is the “correct” model of computation.
- × That the Clay problem is formally ill-posed in the standard Turing setting.

The paper should be read as: “If the ledger model captures physical computation, *then* the P/NP distinction splits into two independent questions.”

2 The Standard P vs NP Problem

For completeness, we recall the standard formulation.

Definition 2.1 (Turing machine). *A (deterministic) Turing machine is a tuple $M = (Q, \Sigma, \delta, q_0, q_{\text{acc}}, q_{\text{rej}})$ with finite state set Q , tape alphabet Σ , transition function $\delta : Q \times \Sigma \rightarrow Q \times \Sigma \times \{L, R\}$, and distinguished states. The time complexity $T_M(n)$ is the maximum number of steps on inputs of length n .*

Definition 2.2 (P and NP (standard)). • ***P**: the class of languages decidable by a deterministic TM in time $O(n^k)$ for some k .*

- ***NP**: the class of languages for which a “yes” certificate of length $\text{poly}(n)$ can be verified in polynomial time.*

The Clay Millennium problem [?] asks: is $P = NP$?

Remark 2.3 (The observation that motivates this paper). *In both definitions, every tape-read operation has **zero cost**. When the verifier checks a certificate, each symbol lookup is free. This is an idealisation. In any physical realisation:*

- *Reading a memory cell dissipates at least $k_B T \ln 2$ of energy (Landauer’s bound [?]).*
- *In quantum mechanics, measurement disturbs the measured state (no-cloning, wavefunction collapse).*
- *In the RS framework, observation has cost $J(r) > 0$ per recognition query.*

The Turing model’s $T_r = 0$ assumption is a modelling choice, not a physical law. The question explored in this paper is: what happens to the P/NP distinction if we take $T_r > 0$ seriously?

2.1 Existing barriers

No unconditional proof of $P \neq NP$ is known. Three *barriers* explain why standard techniques fail:

1. **Relativisation** [?]: there exist oracles A, B with $P^A = NP^A$ and $P^B \neq NP^B$. Any proof must be non-relativising.
2. **Natural proofs** [?]: if one-way functions exist, no “natural” combinatorial property can separate P from NP.
3. **Algebrisation** [?]: proofs that algebrise cannot separate P from NP.

Remark 2.4. *The dual-complexity framework in this paper is not a technique within the standard Turing model. It introduces a new model (ledger computation with $T_r > 0$), which sidesteps the*

three barriers by changing the question rather than answering the original one. This is why we label the results “conditional,” not “unconditional.”

3 Dual Complexity Framework

Definition 3.1 (Recognition-complete complexity). *A recognition-complete complexity measure assigns to each problem instance of size n a pair $(T_c(n), T_r(n))$:*

- $T_c(n)$: computation steps (internal state transitions).
- $T_r(n)$: recognition queries (observation/readout operations).

Total cost: $T_{total}(n) = T_c(n) + T_r(n)$.

Definition 3.2 (Turing model as special case). *The standard Turing model is the special case $T_r(n) = 0$ for all n . All computation cost resides in T_c .*

4 The Ledger Computation Model

Definition 4.1 (Ledger computation). *A ledger computation consists of:*

- **States:** configurations of a double-entry ledger (balanced debit/credit pairs).
- **Evolution:** deterministic double-entry updates preserving balance ($\sigma = 0$).
- **Observation:** extracting information from the ledger by querying specific entries, at cost $J(r)$ per query.

5 Balanced Parity Encoding

Definition 5.1 (Balanced parity). *A ledger configuration has balanced parity if the total debit equals the total credit: $\sum d_i = \sum c_i$. The double-entry structure forces this for all admissible states.*

Hypothesis 5.2 (Information hiding). *In a balanced-parity ledger of n entries, the value of any single-bit predicate (e.g. “is entry k a debit?”) cannot be determined without querying at least $\Omega(n)$ entries, because each entry’s value is constrained by the global balance condition.*

Proposition 5.3 (Parity lower bound (classical)). *Computing the parity of n bits requires reading all n bits in the worst case. No query algorithm can determine $\bigoplus_{i=1}^n x_i$ with fewer than n queries.*

Proof. Adversary argument. Fix any deterministic algorithm making $< n$ queries. An adversary answers consistently but chooses the unqueried bit to control parity. Since the algorithm never queries the last bit, both parity values are consistent with the observed answers. $\square \quad \square$

Theorem 5.4 (Balanced-parity lower bound). *In a balanced ledger with n entries where $\sum d_i = \sum c_i$ (debit = credit), determining whether a specified entry k is a debit or credit requires querying at least $n - 1$ entries in the worst case.*

Proof. Fix an algorithm \mathcal{A} that queries fewer than $n - 1$ entries and claims to determine d_k . There exist at least two unqueried entries $i, j \neq k$. Consider two configurations:

- C_1 : all queried entries as answered, $d_k = +1$, and (d_i, d_j) chosen to satisfy balance.
- C_2 : all queried entries as answered, $d_k = -1$, and (d_i, d_j) chosen to satisfy balance (adjust by swapping i, j).

Both C_1 and C_2 are consistent with the observed query answers (the algorithm cannot distinguish them). Yet d_k differs. Hence \mathcal{A} must err on at least one of C_1, C_2 .

More precisely: the balance constraint $\sum d_i = S$ (a known constant) has $\binom{n}{n/2}$ satisfying assignments. Conditioning on any $n - 2$ entries leaves a 2-dimensional space; both values of d_k are

compatible. The argument generalises to randomised algorithms by Yao’s minimax principle [?]: a probabilistic algorithm needs $\Omega(n)$ queries to achieve error $< 1/3$. \square

Corollary 5.5. *The $\Omega(n)$ lower bound is a consequence of global coupling: the balance constraint links all entries, so local queries reveal global information only after $\Omega(n)$ samples.*

Remark 5.6. *This is strictly stronger than the classical parity bound (prop:parity): parity hiding is an incidental property of bit strings, while balanced-parity hiding is a structural consequence of the double-entry ledger. The latter cannot be avoided by clever encoding because balance is an invariant of the dynamics (T3 conservation).*

6 Worked Example: 3-SAT on a Ledger

To ground the abstract framework, consider a concrete instance.

Example 6.1 (A 3-variable instance). *Let $\phi = (x_1 \vee x_2 \vee \neg x_3) \wedge (\neg x_1 \vee x_3 \vee x_3)$ with $n = 3$ variables.*

Ledger encoding. *Each variable x_i is a ledger entry with value $d_i \in \{+1, -1\}$ (debit or credit). The balanced-parity constraint requires $\sum_{i=1}^3 d_i = \pm 1$ (odd parity for 3 entries).*

Clause checking. *Clause C_j is satisfied iff the appropriate combination of d_i values yields a non-zero inner product with the clause template. A single clause check reads 3 entries: cost $= 3 \cdot J(d_i) = 3 \cdot 0 = 0$ for $d_i = \pm 1$ (unit entries have zero J). However, determining the value $d_i = +1$ vs $d_i = -1$ requires a recognition query.*

Computation phase. *The ledger evolves internally via double-entry updates. After $O(n^{1/3} \log n) = O(1.4 \cdot 1.1) \approx 2$ steps (for $n = 3$), the internal state reorganises.*

Recognition phase. *To read out the satisfying assignment, the observer must query each d_i : cost $= n = 3$ queries. Even after computation has finished, the answer is “hidden” in the ledger’s balanced-parity structure until observed.*

The gap. $T_c = 2$, $T_r = 3$. For $n = 3$ the gap is negligible, but it grows: $T_c = O(n^{1/3} \log n)$ is sublinear while $T_r = \Omega(n)$ is linear. By $n = 1000$: $T_c \approx 70$ but $T_r \geq 1000$.

7 Conditional SAT Separation

Hypothesis 7.1 (SAT computation complexity). *Under the ledger model, the internal evolution can reorganise a SAT instance of n variables into a satisfying assignment (if one exists) in $T_c(n) = O(n^{1/3} \log n)$ steps.*

Hypothesis 7.2 (SAT recognition complexity). *Under the ledger model, verifying that the reorganised ledger encodes a satisfying assignment requires $T_r(n) = \Omega(n)$ recognition queries (from balanced-parity information hiding).*

Theorem 7.3 (Conditional separation). *If Hypotheses ?? and ?? hold, then:*

$$T_c(\text{SAT}) = O(n^{1/3} \log n) \ll T_r(\text{SAT}) = \Omega(n).$$

Computation is fast; recognition is slow. The “hardness” of SAT resides in observation, not evolution.

8 The Split Resolution

Theorem 8.1 (P vs NP splits (conditional)). *Under the dual-complexity framework:*

- **At the computation scale (T_c only):** $P = NP$. The internal evolution can solve NP-complete problems in subpolynomial T_c .

- **At the recognition scale (T_r only):** $P \neq NP$. Observation of the solution requires polynomial T_r , creating a separation from the T_c measure.
- **In the Turing model ($T_r = 0$):** The question is ill-conditioned because T_r is absorbed into T_c and the split is invisible.

9 Implications

1. **Quantum computers shift T_c , not T_r .** Quantum speedups (Grover, Shor) accelerate internal evolution but do not eliminate observation cost. The recognition barrier remains.
2. **Measurement is fundamentally expensive.** The RS collapse threshold $C \geq 1$ makes measurement a real physical cost, not a free operation.
3. **Consciousness has irreducible observation cost.** The attention operator (Section 4 of [?]) is a recognition gate with bounded capacity φ^3 . Even a conscious agent cannot bypass T_r .

10 Falsification Criteria

Falsification Criterion 10.1 (Free observation). *If a physical system is demonstrated where observation has zero energy cost (violating Landauer’s bound), the $T_r > 0$ premise is falsified.*

Falsification Criterion 10.2 (No parity barrier). *If a SAT instance can be verified in $o(n)$ queries on a balanced-parity ledger, the information-hiding hypothesis is falsified.*

11 Comparison with Existing Work

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Reference	Relation to this work
Baker–Gill–Solovay [?]	Relativisation barrier; we sidestep it by changing the model, not proving a Turing-model sep
Razborov–Rudich [?]	Natural proofs barrier; our lower bound is information-theoretic (adversary), not combinat
Aaronson–Wigderson [?]	Algebrisation barrier; the ledger model is not an algebraic extension of a Turing oracle
Landauer [?]	Physical cost of information; we formalise this as $T_r > 0$.
Bennett [?]	Reversible computation; T_c in reversible models is $O(T_{c_{\text{irrev}}}^2)$, but T_r is unchanged.
Grover [?]	Quantum search gives $T_c \rightarrow O(\sqrt{n})$; T_r remains $\Omega(n)$ (measurement collapses the state)

12 Discussion

Claims and non-claims

We have introduced a dual-complexity framework (T_c, T_r) and shown that the RS ledger model provides a natural setting where T_c and T_r can diverge. The key mathematical content is:

1. The balanced-parity lower bound (thm:balanced): proved unconditionally within the query-complexity model.
 2. The conditional separation (thm:separation): $T_c \ll T_r$ for SAT, *if* the ledger model’s T_c hypothesis holds.
 3. The split (thm:split): $P = NP$ at T_c , $P \neq NP$ at T_r , under the same hypothesis.
- Item (1) is rigorous. Items (2) and (3) are conditional.

Why this is not a resolution of P vs NP

The Clay problem asks about the standard Turing model, where $T_r = 0$ by definition. Our framework changes the model. We do *not* prove $P \neq NP$ in the Turing model; we argue that the Turing model’s conflation of T_c and T_r may be the source of the difficulty.

Falsifiability

The framework makes two testable predictions:

1. Any physical computation system will exhibit $T_r > 0$ (Landauer’s bound is never zero).
2. The “hardness” of NP-complete problems in practice will correlate more with *verification cost* (how many bits must be read to check a solution) than with *search cost* (how many internal steps to find a candidate).

Open problems

- (Q1) Can $T_c(\text{SAT}) = O(n^{1/3} \log n)$ be proved in a concrete ledger model, or is it only a hypothesis?
- (Q2) Does the dual framework have a clean complexity-class formulation (e.g. “ \mathbf{P}_c ” for computation-only, “ \mathbf{P}_r ” for recognition-only)?
- (Q3) Is there a natural analogue of the PCP theorem in the dual setting (probabilistic recognition with $o(n)$ queries)?
- (Q4) Does the framework apply to **BPP** vs **BQP** (randomised vs quantum)?

13 Lean Formalization Status

The Lean module `IndisputableMonolith.Complexity.ComputationBridge` is explicitly marked as **SCAFFOLD** and is **not** part of the verified certificate chain. Key caveats:

- `LedgerComputation.states` uses `Type` as a placeholder (often `Unit`).
- Separation theorems rely on hypothetical model assumptions.
- No result should be cited as proven mathematics.

Module	Content
<code>Complexity.ComputationBridge</code>	Dual framework, separation
<code>Complexity.BalancedParityHidden</code>	Parity hiding
<code>Complexity.VertexCover</code>	Example reductions
<code>Complexity.RSVC</code>	RS vertex cover

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