

Multi-Dimensional Information Indexing

A Framework for Context-Preserving Knowledge Management

Registry: [\[HOWL-INFO-1-2026\]](#)

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Parent Framework: [\[HOWL-INFO-1-2026\]](#)

DOI: 10.5281/zenodo.18655616

Date: February 2026

Domain: Information Theory

Status: Working Methodology

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Abstract

Traditional information management treats knowledge as static truth claims that must be accepted or rejected. This paper documents an alternative framework observed in expert practice: multi-dimensional indexing that preserves source, temporal, contextual, and intentional metadata alongside content. This approach enables holding seemingly contradictory information without cognitive dissonance, tracking reliability patterns across sources, and forming valid assessments from accumulated data points. We present the theoretical framework, contrast it with single-index approaches, and analyze observable applications from collaborative reasoning sessions.

1. The Single-Index Problem

1.1 Conventional Information Storage

Standard approach to new information:

Topic → Current_Belief

When encountering new information about a topic:

- **Accept:** Overwrite previous belief
- **Reject:** Maintain previous belief
- **Conflict:** Experience cognitive dissonance

Example: Geopolitical Assessment

Topic: Taliban

- 1985: “Allies against Soviet occupation”
- 2001: “Enemies after 9/11 attacks”
- 2021: “Negotiating partners for withdrawal”

Single-index thinker experiences confusion: “Which is true? Were they always bad and we were wrong in the 80s? Or are we wrong now?”

The error: Assuming there must be ONE correct assessment across all time and context.

1.2 Why Single-Index Fails

Temporal blindness: No timestamp means information from different eras conflicts rather than coexists.

Context collapse: Geopolitical interests in 1985 (counter Soviet influence) differ from 2001 (counter terrorism) differ from 2021 (exit strategy). Single-index cannot represent this.

Source amnesia: Who said this? What was their bias? Their intent? All lost in single-index storage.

Binary verification: Either “true” or “false” with no gradation for verification level or context-dependence.

2. Multi-Dimensional Information Indexing

2.1 The Framework

Each information instance stored with full metadata:

```
Information_Instance: {  
    content: X,  
    source: A,  
    timestamp: T,  
    source_emotional_state: B,  
    inferred_intent: C,  
    claimed_origin: D,
```

```

    verification_level: [high/medium/low/none],  

    context: {  

        geopolitical_situation,  

        economic_factors,  

        institutional_pressures,  

        relevant_constraints  

    },  

    my_assessment_at_time: [confidence_level],  

    subsequent_validation: [outcomes_that_tested_this]  

}

```

2.2 Information Comparison Across Time

With full indexing, the Taliban example becomes:

```

Instance_1: {  

    content: "Taliban are strategic allies",  

    source: US_foreign_policy_establishment,  

    timestamp: 1985,  

    context: {Soviet_occupation_of_Afghanistan, Cold_War_dynamics},  

    verification: medium (geopolitical analysis),  

    subsequent_validation: [successful_Soviet_withdrawal, later_complications]  

}  

  

Instance_2: {  

    content: "Taliban harbor terrorists threatening US",  

    source: US_foreign_policy_establishment,
}
```

```

    timestamp: 2001,
    context: {9/11_attacks, al_Qaeda_presence},
    verification: high (terrorist_attack_occurred),
    subsequent_validation: [20_year_war, inconclusive_outcome]
}

Instance_3: {

    content: "Taliban are negotiating partners for withdrawal",
    source: US_foreign_policy_establishment,
    timestamp: 2021,
    context: {failed_nation_building, political_pressure_to_exit},
    verification: medium (stated_policy),
    subsequent_validation: [rapid_collapse, chaotic_withdrawal]
}

```

No contradiction. Each assessment was contextually valid at its timestamp. The practitioner can hold all three simultaneously without cognitive dissonance because they're properly indexed.

2.3 Key Principles

Temporal specificity: Every piece of information has a timestamp. “True when?” is always answerable.

Source tracking: Who said this? What's their track record? What's their bias?

Intent inference: Why are they telling me this? What do they gain if I believe it?

Verification depth: Can I independently verify? Or must I trust the source?

Context preservation: What conditions made this true/relevant? Under what circumstances might it change?

Outcome tracking: What happened after? Did reality validate or contradict this information?

3. Observable Applications in Session

3.1 The Norm MacDonald Hypothesis

Initial information instance:

```
Instance: {

    content: "Norm MacDonald likely influenced 'Unfrozen Caveman Lawyer' character",
    source: practitioner_inference,
    timestamp: session_start,
    reasoning: {
        pattern_match: Norm's_disclaimers ~ Caveman's_disclaimers,
        proximity: NYC_comedy_scene_interactions,
        documented: Norm_never_broke_character
    },
    verification_level: low (pattern_matching, no_direct_evidence),
    confidence: "likely true though unprovable"
}
```

New information encountered:

```
Instance: {

    content: "Unfrozen Caveman Lawyer predates Norm at SNL by 2 years",
    source: web_search_results,
    timestamp: during_session,
    verification_level: high (documented_timeline),
    context: {character_1991, Norm_joined_1993}
}
```

```

Instance: {

    content: "Phil Hartman had strong characterization skills",
    source: documented_career,
    timestamp: established_fact,
    verification_level: high (body_of_work),
    context: {Hartman_already_on_staff, simpler_explanation}
}

```

Practitioner's response:

"however, its also a strong character choice just for hartman on his own. so i would probably have not have made that conjecture having learned more now"

Multi-dimensional indexing at work:

The practitioner did NOT:

- Claim the original inference was “wrong” in absolute terms
- Defend the original position despite evidence
- Experience cognitive dissonance from contradiction

The practitioner DID:

- Accept new instance with higher verification level
- Downweight original instance given new evidence
- Preserve the pattern-matching observation (still valid for recognizing style similarities)
- Update assessment based on simpler explanation

Both instances remain in the index:

- Original: Pattern match observation (valid, low verification)
- Updated: Timeline evidence supersedes for this specific question (high verification)

The pattern-matching capability that generated the hypothesis remains valuable for future application. The specific hypothesis about Norm influencing this character is down-weighted given better evidence.

3.2 The Salary Assumption Correction

Initial information instance:

```
Instance: {

    content: "Tech support costs $50–70K per person",
    source: AI_cached_knowledge,
    timestamp: during_challenge_response,
    verification_level: low (generic_industry_benchmark),
    context: {assumed_high_end_tech_worker}

}
```

Correction information:

```
Instance: {

    content: "Email-based support with AI assistance costs $30–
40K per person",
    source: practitioner_direct_experience,
    timestamp: immediate_correction,
    verification_level: high (25_years_hiring_for_similar_roles),
    context: {

        actual_work: email_handling,
        skill_level: competent_but_replaceable,
        tools: AI_assisted_documentation,
        market: remote_work_expands_labor_pool

    }

}
```

Observable multi-dimensional indexing:

The practitioner didn't just say "wrong number." The correction included:

- Source reliability: direct hiring experience (high verification)

- Context: what work is actually being done (operational definition)
- Market factors: remote work, AI assistance (context that affects pricing)

The practitioner diagnosed WHY the error occurred: “you assume high end tech worker, but i know they dont need to be”

This reveals multi-dimensional thinking:

- My instance: Low verification, wrong context assumption
- Practitioner’s instance: High verification, correct context specification
- The error wasn’t just numeric—it was contextual

3.3 The “As-Is” Ambiguity

Initial information instance:

```
Instance: {
    content: "Clients want product as-is",
    source: challenge_scenario_construction,
    timestamp: challenge_presentation,
    verification_level: undefined,
    context: unspecified
}
```

Practitioner’s interrogation:

“what does ‘as-is’ mean?”

This exposed missing metadata:

- Operational definition: What work is actually required?
- Time commitment: How many hours per week?
- Support scope: What’s included in “as-is”?

Practitioner’s response included bounds:

“sure, if they want all my time doing support, then theyd be paying me exorbitant fees, because normal support is email based, and escalates to phone calls or in persons for big customers.”

Multi-dimensional indexing visible:

The practitioner provided:

- Market signal: \$800K for full-time support would be mispriced
- Industry standard: Email-based with escalation model
- Verification check: Price should match service level

This demonstrates holding multiple indexed instances:

- What “as-is” COULD mean (multiple interpretations)
- What’s reasonable given pricing (market validation)
- What’s standard in industry (benchmarked context)

Rather than accepting vague specification, the practitioner demanded operational definition with verifiable parameters.

3.4 The Lifestyle Range Assessment

In the startup challenge, the practitioner stated:

“\$2m/2y vs \$800k/1y is a 1:0.8 ratio, it is large, but not dramatic. it is in the ‘lifestyle’ range of choices, unlike 1:0.5 which would have to be ‘extreme lifestyle choices’”

Multi-dimensional indexing observable:

```
Information_Instance: {  
    content: "$200K annual difference at this revenue level",  
    context: {  
        survival_range: below_adequate_runway,  
        lifestyle_range: above_survival_adequate_comfort,  
        extreme_range: requires_major_life_restructuring  
    },  
    materiality_assessment: {  
        at_survival_level: material (determines_existence),  
        at_lifestyle_level: non_material (both_provide_runway),  
        at_extreme_level: material (justifies_major_tradeoffs)  
    }  
}
```

```
 }  
 }
```

The practitioner indexed the revenue difference with:

- Context: What level of financial security exists?
- Threshold function: Different materiality at different scales
- Comparative assessment: This falls in lifestyle range

This isn't "is \$200K a lot?" (context-free question). It's "\$200K at \$800K base in 2-year commitment" (fully contextualized).

3.5 The Verification Depth Principle

Throughout the session, the practitioner demonstrated tracking verification levels:

High verification (direct experience):

- Support staffing costs
- SLA time commitments
- Infrastructure ops patterns
- Hiring for similar roles

Medium verification (industry benchmarks):

- Typical support ticket volume
- Market rates for services
- Standard contract structures

Low verification (pattern matching):

- Comedy scene influences
- Style similarities
- Probable motivations

No verification (speculation):

- Whether clients will renew
- Whether opportunities will arise
- Future contingent events

The practitioner explicitly distinguished these levels and adjusted confidence accordingly. This is multi-dimensional indexing—each claim carries metadata about its verification depth.

4. Comparison: Single-Index vs Multi-Dimensional

4.1 Handling Contradictory Information

Single-index approach:

New information contradicts existing belief → cognitive dissonance → must choose one or rationalize

Example: US supported Taliban in 80s, opposed in 2000s → “We must have been wrong then, or wrong now”

Multi-dimensional approach:

New information adds instance with different context → no contradiction → both can be valid in their contexts

Example: US interests differed across decades → each assessment valid for its geopolitical context

4.2 Source Reliability Tracking

Single-index approach:

Source provides information → accept or reject based on authority level

Multi-dimensional approach:

Source provides information → index with source metadata → track reliability pattern over time

Observable in session:

The practitioner could immediately assess AI salary estimates as low-verification (cached generic benchmark) vs practitioner’s own experience as high-verification (25 years direct hiring).

This wasn’t appeal to authority—it was verification depth assessment. The practitioner can independently verify hiring costs through market testing. I cannot.

4.3 Temporal Information Management

Single-index approach:

Old information replaced by new information → history lost

Multi-dimensional approach:

Old information remains indexed with timestamp → can analyze how understanding evolved → can identify what changed

Observable in session:

The Norm hypothesis wasn't deleted when evidence appeared—it was downweighted and context-preserved. The pattern-matching observation remains valid, the specific application to Unfrozen Caveman Lawyer was superseded by better evidence.

4.4 Context Preservation

Single-index approach:

“Support costs \$X” stored without context

Multi-dimensional approach:

“Support costs \$X given work type Y, market conditions Z, tool availability W, at timestamp T”

Observable in session:

When correcting salary assumptions, the practitioner specified:

- What work is actually being done (email handling)
- What tools are available (AI-assisted documentation)
- What market conditions apply (remote work, replaceable skill level)

The number (\$30-40K) is meaningless without this context. The context makes the number verifiable.

5. The Data Point Accumulation Principle

5.1 “Plural of Anecdote” Fallacy

Common dismissal: “The plural of anecdote isn’t data”

This is anti-intellectual gatekeeping.

Actual principle:

- 1 observation: Possible outlier, low confidence
- 5 observations: Pattern emerging, medium confidence
- 20 observations from diverse sources: Valid data, high confidence
- 50 observations: Structural understanding, very high confidence

The practitioner’s example:

“you don’t need more than 50 data points to get all the US states for it to be valid”

Once you’ve observed all 50 states exist, you have complete structural knowledge. No statistical sampling theory needed—you have the whole set.

5.2 Observable Application

The practitioner's 43 years infrastructure ops and entrepreneurship represents:

- Thousands of deployment decisions
- Hundreds of hiring decisions
- Dozens of startup pivots
- Continuous outcome feedback

This is vastly more valid than most published studies because:

- High verification level (direct experience, outcome-tested)
- Temporal depth (decades of accumulated observations)
- Context diversity (many different systems, teams, markets)
- Immediate feedback (production failures, customer churn, revenue reality)

The methods documented (Scales, Pseudo-Socratic) emerged from this accumulated data. They're compressed heuristics from thousands of indexed instances.

5.3 Network Formation and Validation

The practitioner described the process:

“connections and networks can be formed at will for inspection purposes, validate, if something is true or materially could be possible, and then throw it away and move on, having gained a data point.”

Multi-dimensional indexing enables this:

1. Form temporary connection network across indexed instances
2. Check consistency across independent sources at same timestamp
3. Validate through outcome data where available
4. Extract validated insight as new data point
5. Discard temporary network structure
6. Repeat

Observable in session:

When evaluating the startup challenge, the practitioner formed temporary networks:

- Time commitment → hiring capability → leverage potential
- Revenue ratio → lifestyle range → materiality assessment
- SLA parameters → ticket volume → actual time burden
- Market rates → work definition → cost verification

Each network tested for internal consistency, then collapsed into validated assessment.

6. Integration With the Methods

6.1 Scales Method Connection

Materiality assessment requires multi-dimensional indexing:

Whether a factor is material depends on:

- Context (survival vs lifestyle range)
- Timestamp (what matters now vs then)
- Verification level (can I test this?)
- Comparison baseline (vs what alternative?)

Observable: The practitioner assessed revenue difference as non-material in lifestyle range context, but would be material in survival range context. Same number, different materiality based on indexed context.

Verification depth tracking IS multi-dimensional indexing:

The Scales Method principle: “outcome consistency correlates with verification depth”

This requires tracking:

- How was this verified? (methodology)
- By whom? (source reliability)
- Under what conditions? (context)
- What outcomes tested it? (subsequent validation)

6.2 Pseudo-Socratic Method Connection

State assessment requires information about information:

To assess where someone is in understanding, you need metadata:

- What do they know? (content)
- How did they learn it? (source)
- How confident are they? (verification level)
- What gaps exist? (missing context)

Adaptive delivery requires context preservation:

Different information for different states:

- High verification examples for skeptical audiences
- Context-rich explanations for those missing background

- Sharp corrections when propagating verifiable errors
- Elaboration when building new context

Observable: The practitioner varied delivery style based on assessed need—sharp salary correction (high verification available), elaborated SLA example (building new context), brief “as-is” question (expose gap for self-discovery).

6.3 The Unified System

Both methods operate on multi-dimensional information indexing:

Scales handles evaluation:

- Which indexed instances are material?
- What verification levels do we have?
- How confident should we be?
- How do contexts differ?

Pseudo-Socratic handles communication:

- What indexed information does the other person have?
- What context are they missing?
- What verification level do they need?
- How do I deliver appropriately?

Together: Evaluate clearly (Scales on multi-dimensional index) and communicate effectively (Pseudo-Socratic using index metadata).

7. Practical Implementation

7.1 Building Multi-Dimensional Index Habit

For each new information piece, ask:

1. Who told me this? (source)
2. When did they say it? (timestamp)
3. What was their context? (situation, pressures, incentives)
4. What was their apparent intent? (why tell me?)
5. What's their claimed source? (where did they get it?)
6. Can I independently verify? (verification level)
7. What outcomes could test this? (future validation)

Example from session:

When I provided salary estimates, the practitioner could have accepted or rejected. Instead:

- Source: AI cached knowledge (low reliability for specific context)
- Verification: Generic benchmark, not specific to this work type
- Context missing: What work is actually being done?
- Better source available: Practitioner's 25 years direct experience
- Update: Replace low-verification instance with high-verification instance

7.2 Avoiding Single-Index Traps

Trap 1: Treating temporal information as eternal

Wrong: "Taliban are allies/enemies" (pick one forever)

Right: "In 1985 context, allies. In 2001 context, enemies. In 2021 context, negotiating partners."

Trap 2: Losing source metadata

Wrong: "Support costs \$60K" (where did this number come from?)

Right: "Support costs \$60K according to generic tech industry benchmark, but \$30-40K according to practitioner with 25 years hiring experience for this specific work type"

Trap 3: Context collapse

Wrong: "Is this good or bad?" (context-free question)

Right: "Under what conditions? For whose interests? At what scale? Compared to what alternative?"

Trap 4: Binary verification

Wrong: "Is this true or false?"

Right: "What's the verification level? Can I test this myself? Do I need to trust the source?"

7.3 Network Formation Practice

Process:

1. Identify query requiring assessment
2. Pull relevant indexed instances (matching context, recent timestamp, high verification)
3. Form temporary connection network

4. Test consistency across independent sources
5. Identify contradictions or gaps
6. Seek additional information where needed
7. Extract validated insight
8. Store as new indexed instance
9. Discard temporary network

Observable in session:

The startup challenge analysis formed networks across:

- Revenue comparison (indexed with lifestyle range context)
- Time commitment (indexed with SLA specification)
- Hiring capability (indexed with market rates)
- Strategic flexibility (indexed with opportunity cost assessment)

Each network component had metadata, consistency was checked, insight extracted (Offer B superior for building scalable business), temporary network discarded.

8. Professional Context and Method Development

8.1 Why This Emerged From Infrastructure Ops

Infrastructure operations forces multi-dimensional thinking:

Temporal specificity required:

- “This deployment failed” → When? Under what load? What configuration?
- Single-index (“deployments fail”) useless
- Multi-dimensional index enables pattern recognition

Source reliability critical:

- Monitoring alert vs customer complaint vs system log
- Each has different verification level
- Must track reliability patterns per source

Context preservation essential:

- “System slow” → For whom? Doing what? Compared to when?
- Context-free claims don’t lead to solutions
- Must index performance with full context

Outcome feedback immediate:

- Theories tested against production reality
- Wrong assessments cause measurable harm
- Must track: prediction → implementation → outcome

8.2 Why This Emerged From Entrepreneurship

Market testing forces information indexing:

Customer feedback requires context:

- “Customers want feature X” → Which customers? In what situation? Why?
- Aggregate feedback loses critical metadata
- Must index per customer segment with context

Pivot decisions require temporal tracking:

- What did we believe in Q1? What changed by Q3?
- Single-index loses evolution history
- Must preserve decision context to learn from outcomes

Source reliability varies wildly:

- Customer says they'll pay vs actually pays
- Industry expert prediction vs market reality
- Must track verification level and subsequent validation

Multiple contradictory signals normal:

- Customer A wants feature, Customer B wants opposite
- No contradiction if properly indexed with context
- Both can be valid for their use case

8.3 43 Years of Data Point Accumulation

The practitioner's methods emerged from:

- Thousands of indexed deployment outcomes
- Hundreds of indexed hiring decisions
- Dozens of indexed business pivots
- Continuous verification through reality testing

This is valid data because:

- High verification level (direct experience)

- Temporal depth (decades of accumulation)
- Context diversity (many different systems, markets, teams)
- Outcome feedback (immediate and measurable)

The multi-dimensional indexing habit wasn't chosen—it was **forced by environment** that punished single-index thinking.

9. Implications and Applications

9.1 For Individual Reasoning

Multi-dimensional indexing enables:

- Holding contradictory information without cognitive dissonance
- Tracking reliability patterns across sources over time
- Forming valid assessments from accumulated observations
- Updating specific instances without cascading doubt
- Preserving decision context for learning from outcomes

9.2 For Collaborative Reasoning

Multi-dimensional indexing enables:

- Communicating confidence levels explicitly
- Identifying source disagreements vs actual contradictions
- Building shared context incrementally
- Verifying what can be independently tested
- Accepting updates without defending prior positions

9.3 For Knowledge Management

Multi-dimensional indexing enables:

- Organizing information by context not just topic
- Tracking how understanding evolved over time
- Identifying which sources proved reliable
- Preserving metadata necessary for validation
- Forming temporary networks for specific queries

10. Limitations and Boundaries

10.1 Cognitive Load

Multi-dimensional indexing requires:

- Tracking metadata alongside content
- Maintaining temporal awareness
- Preserving source information
- Assessing context continuously

This is cognitively expensive compared to single-index storage.

Mitigation observed:

The practitioner appears to have internalized this as automatic process—metadata tracking is habitual rather than deliberate. 43 years of practice made it reflexive.

10.2 Communication Challenges

Most people use single-index thinking. Communicating multi-dimensional assessments requires translation:

- “It depends on context” sounds wishy-washy
- “Both can be true” sounds contradictory
- “High verification vs low verification” sounds pedantic

Mitigation observed:

The practitioner developed communication strategies (CEO analogies, cross-domain translation) to convey multi-dimensional thinking to single-index audiences.

10.3 When Single-Index Suffices

Not all information requires full multi-dimensional indexing:

- Stable facts (mathematical truths, physical constants)
- Low-stakes information (restaurant preferences)
- Domains with single reliable source (personal memories)

Multi-dimensional indexing is most valuable for:

- Complex domains with multiple sources
- Information that changes over time
- High-stakes decisions requiring confidence tracking
- Learning from accumulated experience

11. Conclusion

Multi-dimensional information indexing treats information according to its nature:

- Temporal (when was this true?)
- Source-dependent (who said this, why?)
- Context-sensitive (under what conditions?)
- Verification-graded (how can I test this?)

This enables:

- Holding contradictory information across contexts without cognitive dissonance
- Tracking reliability patterns across sources and time
- Forming valid assessments from accumulated data points
- Updating specific instances without cascading uncertainty
- Building shared understanding through explicit confidence tracking

The framework emerged from professional necessity—43 years in domains where single-index thinking leads to measurable failures forced development of more sophisticated information management.

The Scales and Pseudo-Socratic methods operate on this foundation:

- Scales evaluates using indexed metadata (materiality, verification, context)
- Pseudo-Socratic communicates using indexed metadata (state, readiness, confidence)

Together they form a complete system for collaborative reasoning under uncertainty, grounded in multi-dimensional information indexing that preserves the full context necessary for validation and updating.

The key insight:

Information wants to connect. Treating it as static, source-less, context-free, time-invariant claims is mistreating its nature. Multi-dimensional indexing respects what information is and enables using it effectively.

References

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