

1. Formal setup

We start with a single record (or event) as seen by DAIS-10.

- **Raw attribute set:**

Let $A=\{a_1, a_2, \dots, a_n\}$ be the set of attributes present for a record.

- **Raw values:**

Each attribute has a value in some data domain V_i , so a record is:

$$r=(a_1, v_1), \dots, (a_n, v_n)$$

- **Semantic roles (MCM-10 vocabulary):**

Define a finite set of semantic roles

$$R=\{\text{MD}, \text{ME}, \text{MX}, \text{MN}\}$$

where:

- MD = Meaning-Defining
- ME = Meaning-Enhancing
- MX = Meaning-Extending
- MN = Meaning-Neutral
- **Tiers (TIER-10 vocabulary):**

$$T=\{\text{E}, \text{EC}, \text{C}, \text{CN}, \text{N}\}$$

(Essential, Semi-Essential, Contextual, Semi-Contextual, Non-Essential).

- **Semantic importance scale (SICM-10):**

Each attribute will get a semantic score in:

$$S_i \in [0, 100]$$

SIS-10 and SIF-10 will now be defined as **mathematical functions** on these sets.

2. SIS-10: Semantic Interpretation System

2.1 Definition

SIS-10 is a mapping from raw attributes to **semantic descriptors**:

- A semantic descriptor for attribute a_i is:

$$\sigma_i = (r_i, t_i, s_i)$$

where:

- $r_i \in R$ is the semantic role,
- $t_i \in T$ is the tier,
- $s_i \in [0,100]$ is the semantic importance score.

Define:

$$SIS: A \times V \rightarrow \Sigma$$

where $V = \prod_i V_i$, $\Sigma = R \times T \times [0,100]$.

For each (a_i, v_i) in a record:

$$SIS(a_i, v_i) = (r_i, t_i, s_i)$$

We require that SIS-10 is **meaning-driven**, not mechanically driven:

Assumption (Meaning Precedes Mechanics): For any two attributes a_i, a_j with different semantic meanings but identical data types and raw statistics, SIS may assign different (r, t, s) . Formally, SIS is not a function of type or cardinality alone.

2.2 Structural constraints ([link to DAIS-10](#))

We encode the DAIS-10 assumptions as constraints:

1. Continuum constraint:

There exists a function:

$$f_{cont}: T \rightarrow [0,100]$$

that gives a **baseline** importance for each tier, such that:

$$f_{cont}(E) > f_{cont}(EC) > f_{cont}(C) > f_{cont}(CN) > f_{cont}(N)$$

and actual scores satisfy:

$$s_i \approx f_{cont}(t_i) \text{ up to a local adjustment}$$

2. Tier–role compatibility:

For MD attributes, SIS must never assign **low-tier** with **high score** in a contradictory way. One simple version:

If $r_i = \text{MD}$ then:

$t_i \in \{E, EC\}$ and $s_i \geq s_{MD, \min}$

for some constant $S_{MD, \min} \in (50, 100]$.

3. Domain-agnosticism:

SIS does not depend on **domain labels**, only on semantic mappings. Formally, if two domains have an isomorphism between attribute meaning sets, SIS commutes with that mapping.

3. SIF-10: Semantic Influence Framework

SIF-10 takes the semantic descriptors from SIS-10 and produces **influence weights** that drive downstream decisions.

3.1 Definition

For a record, we have:

$$\Sigma(r) = \{\sigma_i = (r_i, t_i, s_i) \mid i=1, \dots, n\}$$

SIF-10 is a mapping:

$$SIF : \Sigma^n \rightarrow [0,1]^n$$

producing influence weights:

$$w_i = SIF(\sigma_i \mid \Sigma(r))$$

We require:

- **Normalization:**

$$\sum_{t \rightarrow 1 \text{ to } n} \underline{W}_i = 1$$

- **Monotonicity in score:** If $s_i > s_j$ and all else equal, then $w_i > w_j$.
- **Tier precedence:** For attributes with different tiers, dominance is preserved.

3.2 Concrete influence function

Define a **base importance function**:

$$g : R \times T \times [0,100] \rightarrow R > 0$$

A simple and powerful form:

$$g(r_i, t_i, s_i) = \alpha(r_i) \cdot \beta(t_i) \cdot h(s_i)$$

Where:

- $\alpha: R \rightarrow R > 0$ encodes **role weight** (MD > ME > MX > MN).
- $\beta: T \rightarrow R > 0$ encodes **tier weight** (E > EC > C > CN > N).
- $h: [0,100] \rightarrow R > 0$ is a **strictly increasing** function of semantic score.

Then define:

$$w_i = g(r_i, t_i, s_i) / \sum_{j=1}^n g(r_j, t_j, s_j)$$

This is SIF-10.

4. Theorems and proofs

Now we prove the key properties that DAIS-10 claims for SIS-10 and SIF-10.

4.1 Theorem 1 — Essential attributes always dominate influence

Statement: Suppose we choose α, β, h such that:

$$\alpha(MD) \geq \alpha(ME) \geq \alpha(MX) \geq \alpha(MN) >$$

$$\beta(E) > \beta(EC) > \beta(C) > \beta(CN) > \beta(N) >$$

and h is strictly increasing. Then any attribute classified as Essential with sufficiently high s will have **greater influence weight** than any Non Essential attribute with lower s .

Proof:

Let $\sigma_i = (r_i, t_i, s_i)$ with $t_i = E$ and $\sigma_j = (r_j, t_j, s_j)$ with $t_j = N$.

Then:

$$g_i = \alpha(r_i) \beta(E) h(s_i)$$

$$g_j = \alpha(r_j) \beta(N) h(s_j)$$

By assumption:

$$\beta(E) > \beta(N)$$

and h is strictly increasing, so for any $s_i > s_j$:

$$h(s_i) > h(s_j)$$

Even if $\alpha(r_i) \leq \alpha(r_j)$, we can choose score thresholds such that:

$$\alpha(r_i) \beta(E) h(s_i) > \alpha(r_j) \beta(N) h(s_j)$$

Thus $g_i > g_j$. Since:

$$W_k = g_k / \sum \ell g_\ell$$

we have $g_i > g_j \Rightarrow w_i > w_j$.

Hence, under this construction, sufficiently important Essential attributes **dominate** Non Essential ones in influence weight. \square

This formalizes the DAIS-10 claim: **tier-weighted governance and influence**.

4.2 Theorem 2 — Semantic continuity (no abrupt jumps)

Statement: If h is continuous and strictly increasing, and SIS-10 ensures that s_i varies continuously along the importance continuum, then SIF-10 influence weights w_i vary **continuously** with s_i . No discrete jumps in influence occur from small changes in semantic score.

Proof:

We consider w_i as a function of all scores s_1, \dots, s_n .

Each base importance term is:

$$g_i(s_i) = \alpha(r_i) \beta(t_i) h(s_i)$$

$\alpha(r_i)$ and $\beta(t_i)$ are constants for fixed role and tier, and h is continuous. Thus $g_i(s_i)$ is continuous in s_i .

The denominator:

$$G(s_1, \dots, s_n) = \sum (j \rightarrow 1 \leq n) g_j(s_j)$$

is a finite sum of continuous functions, hence continuous.

Then:

$$w_i(s_1, \dots, s_n) = g_i(s_i) / G(s_1, \dots, s_n)$$

is a quotient of continuous functions where the denominator is strictly positive (all $g_j > 0$ by construction). Hence w_i is continuous in all s_j .

Thus small changes in semantic scores s_i produce small changes in influence weights w_i , i.e. **no abrupt jumps**. \square

This satisfies the DAIS-10 assumption that importance is a **continuum, not a binary step function**.

4.3 Theorem 3 — Invariance under semantically isomorphic re-labelings

Statement: If two domains D and D' are semantically isomorphic via a bijection ϕ that preserves roles, tiers, and scores, then SIS-10 and SIF-10 produce **equivalent influence structures** up to relabeling.

Setup:

Let domain D have attributes A and domain D' have attributes A' . Suppose there exists a bijection:

$$\phi: A \rightarrow A'$$

such that for all $a_i \in A$:

$$SISD(a_i, v_i) = (r_i, t_i, s_i)$$

$$SISD'(\phi(a_i), v'_i) = (r_i, t_i, s_i)$$

i.e., the semantic descriptors are identical.

Proof:

For a record in D , SIF-10 computes:

$$W_i = g(r_i, t_i, s_i) / \sum_j g(r_j, t_j, s_j)$$

For the corresponding record in D' with attributes $\phi(a_i)$, we get the same set of (r_i, t_i, s_i) , hence the same g_i and same normalized weights w_i .

Thus, up to relabeling indices by ϕ , the influence vector in D and D' is identical.

This proves **domain-agnosticism** at the level of SIS-10 and SIF-10: they depend only on semantic structure, not on domain names. \square

4.4 Theorem 4 — Essential-missing implies semantic collapse (record failure)

Statement: Suppose DAIS-10 defines a record as semantically valid only if all Meaning-Defining Essential attributes are present. Under SIS-10 and SIF-10 as defined, missing such attributes implies that **any reasonable completeness or quality function** will flag the record as failed.

Setup:

Define the set of Essential attributes for a record:

$$E = \{ i \mid t_i = E \wedge r_i = MD \}$$

Define a completeness/validity function:

$$C(r) = \begin{cases} 0 & \text{if } \exists i \in E \text{ s.t. } a_i \text{ is missing} \\ -1A & \end{cases}$$

$$\begin{cases} f(\{w_i\}) & \text{otherwise---1B} \end{cases}$$

for some function f on influence weights.

Proof:

By definition, if any MD+E attribute is missing, we assign $C(r)=0$.

This is not a numerical accident but a **semantic rule**: SIS-10 cannot produce a semantic descriptor for a missing attribute, so SIF-10 cannot assign influence. DAIS-10's governance assumption “Missing Essential attributes = record failure” is implemented by this rule.

Formally, the presence of MD+E attributes is a **precondition** for further scoring. Thus, any valid scoring function consistent with DAIS-10 will treat their absence as semantic collapse. \square

This matches 1.5.2 exactly.

5. What we have now

We've done three things:

1. **Refined SIS-10** as a function that assigns roles, tiers, and continuous semantic scores, constrained by DAIS-10's ontological and semantic assumptions.
2. **Refined SIF-10** as a normalized influence functional using role weights, tier weights, and a strictly increasing, continuous mapping from semantic score to raw importance.
3. **Proved key properties:**
 - o **Tier dominance:** Essential (E) with MD role can be guaranteed to dominate Non-Essential under SIF-10.
 - o **Continuity:** Influence weights change smoothly with semantic importance (no abrupt jumps).
 - o **Domain-agnostic invariance:** Semantically equivalent domains yield identical influence patterns.
 - o **Record failure semantics:** Missing Essential MD attributes mathematically forces record failure as per governance rules.

These are exactly the kinds of **internal consistency theorems** a standards document can include to show DAIS-10 is not just philosophical but **formally coherent**.

MCM-10 — Meaning Classification Model

Purpose

Assigns each attribute a **semantic role** based on its contribution to meaning.

Semantic Roles

$R=\{MD, ME, MX, MN\}$

- **MD — Meaning-Defining** Required for identity, purpose, or interpretability.
- **ME — Meaning-Enhancing** Clarifies or enriches meaning.
- **MX — Meaning-Extending** Adds analytical depth or optional context.
- **MN — Meaning-Neutral** No semantic contribution.

Formal Definition

MCM-10 is a mapping:

$MCM:A \rightarrow R$

For each attribute a_i :

$$r_i = MCM(a_i)$$

Constraints

1. MD attributes must map to Essential tiers (proved later).
2. MN attributes must map to Non-Essential tiers.
3. ME and MX may map to intermediate tiers.

TIER-10 — Tier Assignment System

Purpose

Assigns each attribute a **tier** representing its governance level.

Tiers

$$T = \{E, EC, C, CN, N\}$$

- **E — Essential**
- **EC — Semi-Essential**
- **C — Contextual**
- **CN — Semi-Contextual**
- **N — Non-Essential**

Formal Definition

$$TIER: R \times \text{Context} \rightarrow T$$

For each attribute:

$$t_i = TIER(r_i, \text{context})$$

Constraints

1. If $r_i = MD$, then $t_i \in \{E, EC\}$.
2. If $r_i = MN$, then $t_i = N$.
3. Context may shift ME/MX between EC/C/CN.

SICM-10 — Semantic Intensity Continuum Model

Purpose

Assigns each attribute a **semantic intensity score** on a continuous 0–100 scale.

Formal Definition

$$\text{SICM}: T \times \text{Context} \rightarrow [0, 100]$$

For each attribute:

$$s_i = \text{SICM}(t_i, \text{context})$$

Continuum Constraint

There exists a strictly decreasing function:

$$f_{\text{cont}}(E) > f_{\text{cont}}(EC) > f_{\text{cont}}(C) > f_{\text{cont}}(CN) > f_{\text{cont}}(N)$$

Actual score:

$$s_i = f_{\text{cont}}(t_i) + \epsilon_i$$

where ϵ_i is a contextual adjustment.

Properties

- Continuous
- Monotonic
- Tier-aligned
- Context-sensitive

DIFS-10 — Drift & Fading Subzones

Purpose

Models **semantic decay** over time or under uncertainty.

Subzones

$$Z = \{E1, EC1, C2, CN1, N1\}$$

Each subzone represents a **fading stage** within a tier.

Formal Definition

Let $s_i(t)$ be the semantic score at time t .

DIFS-10 defines:

$$ds_i / dt = -\lambda_i \cdot s_i(t)$$

Where:

- λ_i = drift coefficient
- Higher tiers have lower drift rates:

$$\lambda_E < \lambda_{EC} < \lambda_C < \lambda_{CN} < \lambda_N$$

Subzone Boundaries

Each tier has internal thresholds:

$$E1:s_i \in [80, 100]$$

$$EC1:s_i \in [60, 80)$$

$$C2:s_i \in [40, 60)$$

$$CN1:s_i \in [20, 40)$$

$$N1:s_i \in [0, 20)$$

Properties

- Drift is exponential
- Fading is smooth
- Subzones are continuous intervals

QFIM-10 — Qualified Interpretation Model

Purpose

Produces a **qualified interpretation** of meaning based on score, tier, and drift.

Interpretation Levels

$Q = \{\text{Critical}, \text{High}, \text{Moderate}, \text{Low}, \text{Minimal}\}$

Formal Definition

$\text{QFIM} : (s_i, t_i, z_i) \rightarrow Q$

Where:

- s_i = semantic score
- t_i = tier
- z_i = subzone

Mapping Rule

Define thresholds:

Critical : $s_i \geq 90$

High : $70 \leq s_i < 90$

Moderate : $50 \leq s_i < 70$

Low : $30 \leq s_i < 50$

Minimal : $s_i < 30$

Tier and subzone adjust thresholds upward or downward.

AMD-10 — Automated Meaning Diagnostics

Purpose

Detects semantic failures, contradictions, and anomalies.

Diagnostic Classes

$D = \{\text{Critical}, \text{Major}, \text{Minor}, \text{None}\}$

Formal Definition

$\text{AMD} : \Sigma^n \rightarrow D$

Where Σ is the set of semantic descriptors.

Rules

1. **Critical Failure** Missing any MD+E attribute:

$\exists i : r_i = \text{MD} \wedge t_i = \text{E} \wedge \text{missing}(a_i)$

2. **Major Failure** Contradictions among Meaning-Enhancing attributes:

$\exists i, j : r_i = r_j = \text{ME} \wedge \text{contradict}(a_i, a_j)$

3. **Minor Failure** Missing Contextual attributes:

$r_i = \text{MX} \wedge \text{missing}(a_i)$

4. **No Failure** All semantic constraints satisfied.