

# 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

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

where:

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

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

(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  $s_i \geq s_{\text{MD}, \min}$

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

### 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:

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

producing influence weights:

$w_i = \text{SIF}(\sigma_i \mid \Sigma(r))$

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:

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

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  $\alpha, \beta, h$  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:

$$G(s_1, \dots, s_n) = \sum_{j=1}^n g_j(s_j)$$

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  $\phi$  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  $a_i \in A$ :

$$\text{SISD}(a_i, v_i) = (r_i, t_i, s_i)$$

$$\text{SISD}'(\phi(a_i), v_i') = (r_i, t_i, s_i)$$

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  $\phi$ , 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:

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

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

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:**
  - **Tier dominance:** Essential (E) with MD role can be guaranteed to dominate Non-Essential under SIF-10.
  - **Continuity:** Influence weights change smoothly with semantic importance (no abrupt jumps).
  - **Domain-agnostic invariance:** Semantically equivalent domains yield identical influence patterns.
  - **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  $a_i$ :

$r_i = \text{MCM}(a_i)$

## 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

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

For each attribute:

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

### Constraints

1. If  $r_i = \text{MD}$ , then  $t_i \in \{E, EC\}$ .
2. If  $r_i = \text{MN}$ , 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

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

For each attribute:

$si = SICM(ti, context)$

### Continuum Constraint

There exists a strictly decreasing function:

$fcont(E) > fcont(EC) > fcont(C) > fcont(CN) > fcont(N)$

Actual score:

$si = fcont(ti) + \epsilon_i$

where  $\epsilon_i$  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:

- $\lambda_i$  = 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

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

## Formal Definition

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

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

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

## Formal Definition

$$AMD : \Sigma^n \rightarrow D$$

Where  $\Sigma$  is the set of semantic descriptors.

## Rules

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

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

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

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

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

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

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