

# Signal Theory: The Architecture of Optimal Intent Encoding in Communication Systems

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## Abstract

Every act of communication—from a CEO directing a team to a nervous system firing an impulse to a server routing a packet—follows the same architecture: intent is encoded into a signal, transmitted through a channel, decoded at a destination, and produces action. Despite this universality, no single discipline has produced a unified framework for the fundamental unit of actionable communication. Information theory provides the mathematics of transmission but ignores social convention. Linguistics describes structure but not channels. Speech act theory classifies what utterances do but not how they should be built. Organizational theory models coordination but lacks a rigorous model of what flows through it. Each discipline captures a real dimension of the problem; none captures the whole.

This paper introduces **Signal Theory**, a framework that integrates seven foundational traditions—Shannon’s information theory (1948), Wiener’s cybernetics (1948), Ashby’s Law of Requisite Variety (1956), Beer’s Viable System Model (1972), Bakhtin’s speech genres (1986), Searle’s speech acts (1976), and Kress & van Leeuwen’s multimodal semiotics (2001)—into a single architecture for optimal communication. It defines the **Signal** as the fundamental unit of actionable communication, classified across five dimensions (Mode, Genre, Type, Format, Structure), and the **Optimal System** as a cybernetic architecture organized around a single root objective—maximizing Signal-to-Noise Ratio—constrained by four governing principles that establish the boundary conditions for viable communication. The framework provides eleven distinct failure modes with corrective actions, formal measurement instruments including a requisite variety ratio and triple-layer signal verification, and a technology-independent architecture that applies recursively from a single message to an entire organization.

**Keywords:** signal theory, information encoding, communication architecture, cybernetics, signal-to-noise ratio, viable systems, speech genres, multimodal semiotics, organizational communication, channel capacity

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# 1 Introduction

Communication is the substrate of coordination, yet no unified theory explains why it fails or how to diagnose failure systematically. Teams misalign, directives dissolve between sender and receiver, tools multiply channels without understanding capacity, and when communication fails, practitioners lack a diagnostic framework to identify *which part broke*. The problem is not technological. The problem is that the tools were built without a unified theory, and the theory spanning the relevant disciplines was never integrated.

This paper provides that integration: a unified diagnostic framework organized around a single root objective—**maximize Signal-to-Noise Ratio**—constrained by four governing principles. If communication is failing, one of these four constraints is violated, and each violation has a specific diagnosis and corrective action.

The theoretical foundation begins with Claude Shannon, who stated in 1948 that the fundamental problem of communication is “that of reproducing at one point either exactly or approximately a message selected at another point.” This problem is universal—it applies to telecommunications, to software systems, to biological organisms, and to every organization that has ever tried to coordinate human action through encoded intent. But the fields that study it remain fragmented: information theory, linguistics, speech act theory, organizational theory each capturing a real dimension, none capturing the whole.

There is a deeper problem beneath the fragmentation. Communication does not happen in a vacuum. It happens in a **multidimensional environment that is constantly changing**—where channel noise fluctuates, receiver attention shifts, organizational context evolves, competing signals multiply, and genre conventions drift—and in almost every case, **these changes are not caused by the encoder**. The sender encodes a message into a landscape they did not design, cannot fully observe, and do not control. The question that motivates this paper is:

*How does one encode a message in a multidimensional environment where things are constantly changing but are never changed by them?*

The answer, this paper argues, is the **Signal**—and the architecture that keeps Signal-to-Noise Ratio maximized across shifting conditions is the **Optimal System**.

This paper presents a unifying framework—**Signal Theory**—that integrates these fragmented perspectives into a single, coherent architecture. The framework is organized around one root metric—**Signal-to-Noise Ratio**—and four governing principles that define the constraints on its maximization: channel capacity (Shannon), requisite variety (Ashby), recursive viable architecture (Beer), and closed-loop feedback (Wiener). S/N is not one metric among many; it is the fundamental measure of communication quality. The four principles are the constraints that govern how that quality is achieved. The central contributions are:

1. **The Signal**—a formal definition of the fundamental unit of actionable communication, classified across five dimensions (Mode, Genre, Type, Format, Structure)
2. **The Optimal System**—a cybernetic architecture that satisfies all four governing constraints simultaneously, achieving maximum Signal-to-Noise Ratio across all communication channels at every level of recursion

3. **The Path of Least Resistance**—an argument, grounded in Shannon’s channel coding theorem, Zipf’s Principle of Least Effort, and the empirical evidence of Morse code, that optimal encoding converges on the minimum-effort path through noise
4. **A unified taxonomy** that resolves the naming failure across disciplines—showing that “text,” “document,” “artifact,” “node,” and “textual unit” each describe a single dimension of the multidimensional Signal

The Optimal System (OS) is not an “Operating System” in the computing sense. It is the system configuration that maximizes Signal-to-Noise Ratio—the architecture where all four governing constraints are satisfied simultaneously, at every level of recursion. Because the environment changes continuously, the Optimal System is an asymptotic target—a system that adapts toward maximum S/N, never a static end-state.

Together, these contributions provide the first unified diagnostic for communication failure—classifying where a system fails, which principle it violates, and what corrective action to take. The architecture applies recursively from a single message to an entire organization. Four companion documents extend the theory into applied specifications: *The Optimal System: Architecture Specification*, *Signal Theory: Complete Taxonomy Reference*, *Signal Theory: Signal Composition*, and *Signal Theory: Complete Genre Reference*.

The diagnostic reduces to a single question applied four times: *Is the channel adequate? Is the repertoire sufficient? Is the architecture complete? Is the feedback loop closed?* Any “no” identifies both the failure and its fix. Appendix D walks through a complete diagnostic application.

This paper is itself a Signal. If the framework cannot classify its own transmission, it fails its own test:

S = (M, G, T, F, W)	
M =	Linguistic + Visual (written text with tables, diagrams, and code blocks)
G =	Foundational Paper (academic theoretical framework)
T =	Inform + Direct (states the theory; compels the reader to evaluate it)
F =	Document (PDF / DOCX / Markdown)
W =	Abstract -> Foundations -> Definition -> Taxonomy -> Encoding -> Network -> Composition -> Architecture -> Machine Intelligence -> Path of Least Resistance -> Discussion -> Future Work -> Conclusion
Source:	The author (encoder of the theory)
Channel:	The publication medium through which this document reaches the reader
Destination:	Leaders, engineers, designers, and researchers who have watched communication fail and need to know which part broke and how to fix it
Noise:	Disciplinary jargon gaps, document length, any mismatch between the author’s genre conventions and the reader’s
Feedback:	Peer review, citation, critique, adoption -- or rejection

Every element of Signal Theory applies to Signal Theory. The framework is recursive in its own application.

**Reading paths:** Readers seeking the applied diagnostic can proceed directly from this Introduction to Section III (The Signal), Section VI §6.4 (Failure Modes in the Network), and Appendix D (Diagnostic Walkthrough). Section II provides the theoretical foundations—the proof behind the diagnostic—for readers who want to understand *why* these four constraints hold. The remaining sections build the complete architecture.

## 2 Foundations and Governing Principles

Signal Theory draws on seven foundational bodies of work. Four provide governing principles—two are mathematical theorems with formal proofs (Shannon, Ashby), two are systems-theoretic principles with strong empirical and theoretical support (Beer, Wiener). They differ in epistemological status—the mathematical theorems define hard limits that cannot be exceeded, while the systems-theoretic principles define architectural requirements whose violation produces observable failure—but together they constitute the constraint space within which any viable communication system must operate. Three additional traditions provide the semiotic dimensions that complete the classification system (Bakhtin, Searle, Kress & van Leeuwen).

### 2.1 Shannon: The Mathematical Theory of Communication (1948)

Claude Shannon’s “A Mathematical Theory of Communication” (Shannon, 1948) defined the universal architecture of all communication:

Source -> Encoder -> [Channel + Noise] -> Decoder -> Destination
--

Shannon proved that channel capacity—the maximum rate of reliable communication—is:

$$C = B \times \log_2(1 + S/N) \quad (1)$$

Where  $C$  = Channel Capacity,  $B$  = Bandwidth, and  $S/N$  = Signal-to-Noise Ratio. The implication is absolute: reliable communication above channel capacity is impossible. Shannon further proved that optimal encoding schemes exist—codes that approach channel capacity with arbitrarily small error probability—establishing that the *quality of encoding* determines how closely a system can approach its theoretical limit.

**Governing Principle (Law 1):** For any communication channel with capacity  $C$ , if the information rate  $R < C$ , there exists an encoding scheme that achieves arbitrarily small error probability. If  $R > C$ , reliable communication is impossible. The system must know the capacity of each channel, encode Signals at rates below that capacity, and use proper encoding (the right Genre, Structure, Mode) to approach the limit. No system can exceed this limit. It can only optimize within it.

**Limitation:** Shannon’s model is channel-focused and content-agnostic. It provides the mathematics of transmission but says nothing about *what* is being transmitted, *how* it should be structured, or *what conventions* the sender and receiver must share to achieve understanding.

This limitation is fundamental, not incidental. Information can be viewed from two irreducible perspectives: **syntactic** and **semantic**. The syntactic aspect—the volume

of information, the bits transmitted, the statistical properties of the message—is exactly what Shannon’s theory measures. Shannon himself was explicit about this scope: “The fundamental problem of communication is that of reproducing at one point either exactly or approximately a message selected at another point. Frequently the messages have *meaning* . . . these semantic aspects of communication are irrelevant to the engineering problem.” Shannon was not denying the existence of meaning. He was defining the boundary of his theorem.

But communication systems—the systems this paper formalizes—cannot stop at syntax. The semantic aspect of information is where meaning resides, and meaning is what provokes action. A perfectly transmitted signal that conveys no meaning to its receiver has achieved Shannon’s objective (faithful reproduction) while failing the communication objective (productive response). This is why Signal Theory requires seven traditions rather than one: Shannon provides the channel mathematics, but Bakhtin provides genre conventions (the semantic contract between encoder and decoder), Searle provides illocutionary force (the action-provoking dimension of signals), and the remaining traditions supply the other dimensions that syntax alone cannot capture.

The relationship between information, knowledge, and action is directional:

- **Information** is the flow of messages—syntactic and semantic combined. It makes visible previously invisible meanings and sheds light on unexpected connections.
- **Knowledge** is created by that very flow of information. Knowledge is not information stored; it is information processed through belief and commitment—the receiver’s existing framework (belief) combined with the receiver’s readiness to act (commitment).
- **Action** follows from knowledge when the context-specific and relational meaning of the information is understood. New information, once it passes through the belief-commitment filter, provokes the receiver into action to accomplish an objective.

The implication for Signal Theory is precise: the Signal  $S = (M, G, T, F, W)$  must be designed to satisfy both syntactic and semantic requirements simultaneously. The syntactic dimension (Mode, Format) ensures the signal can be transmitted within channel capacity. The semantic dimension (Genre, Type, Structure) ensures the signal carries meaning that the receiver’s belief-commitment framework can process into action. A signal optimized only for syntactic fidelity (perfect transmission, no meaning) fails. A signal optimized only for semantic richness (profound meaning, poor encoding) fails. The Optimal System (Section VIII) achieves both: maximum S/N in the syntactic domain *and* maximum actionability in the semantic domain.

This dual requirement—syntactic fidelity and semantic actionability—is why Signal-to-Noise Ratio serves as the root *orientation* of the framework rather than either transmission accuracy or meaning density alone. In digital and telecommunications channels, S/N is a computable quantity with a precise mathematical relationship to channel capacity. In organizational and cognitive channels, S/N is a *directional constraint*—the principle that communication quality is the ratio of actionable intent to interference, that every channel has a ceiling, and that encoding quality determines proximity to that ceiling. The framework does not claim that S/N is computable in organizational contexts; it claims that the *structural principle* S/N represents—maximize signal, minimize noise—is the organizing

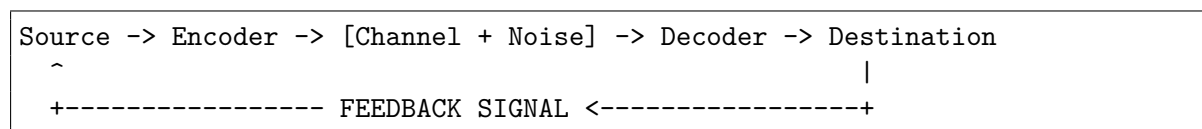
objective, operationalized through the six proxy metrics defined in Section 11.3. This distinction between quantitative and structural application is maintained throughout.

S/N captures both dimensions: signal strength includes semantic content (information that provokes productive action), and noise includes both channel noise (Shannon’s thermal and interference noise) and semantic noise (information that is faithfully transmitted but carries no meaning for the receiver—the wrong genre, the wrong type, the wrong structure for the destination’s belief-commitment framework).

**A note on scope of application:** This paper applies Shannon’s results in two ways: *quantitatively* to channels where  $B$  and  $S/N$  can be measured (telecommunications, software systems, digital protocols), and *structurally* to channels where they cannot (organizational communication, cognitive processing). In the structural application, channel capacity is a directional constraint—the principle that every channel has a finite ceiling and that encoding quality determines how closely a system approaches it—rather than a computable quantity. The equation  $C = B \times \log_2(1 + S/N)$  is not claimed to be computable in organizational contexts. It is the *theorem from which the structural principle derives*: every channel has finite capacity, encoding quality determines how closely capacity is approached, and exceeding capacity guarantees failure. The proxy metrics defined in Section 11.3 operationalize this principle for organizational channels where  $B$  and  $S/N$  cannot be directly measured. Both applications—quantitative and structural—are valid; the distinction matters for measurement (Section 11.3) and is maintained throughout.

## 2.2 Wiener: Cybernetics and the Feedback Loop (1948)

Norbert Wiener, working alongside Shannon at MIT, defined cybernetics as “the study of control and communication in the animal and the machine” (Wiener, 1948). His central insight was **circular causality**—the output of a system feeds back as input, creating a self-correcting loop:



**Governing Principle (Principle 4):** A system cannot self-correct without feedback. Goal-directed behavior requires a closed causal loop where the output is measured against the goal and the difference (error signal) drives corrective action. Linear communication (broadcast) cannot achieve optimal fidelity. Every Signal must have a feedback path. Encoding quality can only improve through iterative error correction—the source must receive information about how the destination decoded the Signal and whether the intended action was achieved.

## 2.3 Ashby: The Law of Requisite Variety (1956)

W. Ross Ashby formulated the First Law of Cybernetics (Ashby, 1956):

**“Only variety can absorb variety.”**

Formally:  $V(C) \geq V(D)$ —the variety of the controller must be at least as great as the variety of the disturbances it faces.

**Governing Principle (Law 2):** The system’s Signal repertoire (genres, modes, formats, structures) must have enough variety to handle every type of situation the environment produces. If the variety of the controller falls short, the unhandled situations become unmanaged disturbances. A system with three communication patterns (email, meeting, Slack message) attempting to handle fifty different situations requiring distinct genres will systematically fail in the situations that exceed its repertoire. The corollary: if the controller’s variety cannot be increased, the variety of the disturbances must be attenuated—narrowing scope, reducing complexity, standardizing operations.

## 2.4 Beer: The Viable System Model (1972)

Stafford Beer, a student of Ashby, applied cybernetics to organizational design (Beer, 1972). His Viable System Model (VSM) defines the architecture that any viable system must have—from a single team to an entire economy. It is recursive: every subsystem is itself a viable system with the same structure. Beer modeled the VSM on the human nervous system, defining five subsystems: Operations (System 1), Coordination (System 2), Control (System 3, plus System 3\* for audit), Intelligence (System 4), and Policy (System 5).

**Governing Principle (Principle 3):** Any viable system requires exactly five subsystems, and each operational subsystem must itself be a viable system with the same five-part structure. Removing System 4 (Intelligence) eliminates the capacity to detect environmental changes. Removing System 2 (Coordination) means operational units will conflict. The five-system structure is the minimum viable architecture, and it must be recursive—each operational unit at every scale must contain all five subsystems.

## 2.5 Bakhtin: Speech Genres (1986)

Mikhail Bakhtin’s “The Problem of Speech Genres” (Bakhtin, 1986) established that every sphere of human activity develops its own “relatively stable types of utterances”—**speech genres**. A genre is not a format or a template. It is a socially recognized pattern of communication that both sender and receiver understand—its thematic content, its compositional structure, and its style are all conventionalized.

**Contribution to Signal Theory:** The Genre dimension. Bakhtin’s insight that genre competence is a skill—“when hearing others’ speech, we guess its genre from the very first words”—explains why a perfectly clear message can fail if sent in a genre the receiver does not recognize.

## 2.6 Searle: The Taxonomy of Illocutionary Acts (1976)

John Searle (Searle, 1976), building on J.L. Austin’s *How to Do Things with Words* (Austin, 1962), classified the communicative functions that utterances perform—what speech acts *do* to the world. Every utterance is either a directive (compels action), an assertive (states facts), a commissive (commits the sender), a declarative (changes reality by being stated), or an expressive (conveys attitude). This framework adopts practitioner labels—**Direct, Inform, Commit, Decide, Express**—that map one-to-one to Searle’s categories while removing the terminological barrier.

**Contribution to Signal Theory:** The Type dimension. Searle’s insight that the same words can perform different acts depending on context—“It’s cold in here” can be an assertive, a directive, or an expressive—establishes that intent, not content, determines communicative function.

## 2.7 Kress & van Leeuwen: Multimodal Semiotics (2001)

Gunther Kress and Theo van Leeuwen’s work on multimodal discourse (Kress and van Leeuwen, 2001) established that meaning is made across multiple perceptual channels—not just language. Written text operates by the logic of *time* (sequential). Visual images operate by the logic of *space* (simultaneous arrangement). Each mode has its own affordances and constraints.

**Contribution to Signal Theory:** The Mode dimension and the principle that choosing the wrong mode is choosing the wrong logic for the message.

## 2.8 The Gap

Each of these bodies of work captures a real dimension of communication:

Theory	What It Describes	What It Misses
Shannon	Channel mathematics	Content, convention, structure
Wiener	Feedback and self-correction	What is being corrected, how it is structured
Ashby	Variety requirements	What the varieties are, how to classify them
Beer	Organizational architecture	The content that flows through the architecture
Bakhtin	Conventionalized forms	Mathematical constraints, channel properties
Searle	Communicative functions	Mode, format, structure, channel
Kress & van Leeuwen	Perceptual channels	Function, convention, mathematical limits

**Table 1:** Foundational theories and their respective gaps.

No existing framework integrates all seven perspectives into a unified model of communication that is simultaneously mathematically constrained (Shannon, Ashby), cybernetically structured (Wiener, Beer), and semiotically complete (Bakhtin, Searle, Kress & van Leeuwen).

There is a further gap—not between disciplines, but between theory and practice. **Information theory has been divorced from information technology.** The field

that created the mathematical foundations of communication has remained in telecommunications. The fields that build communication systems—software engineering, organizational design, product management—build without the theory. The result: organizations build tools (Slack, Notion, Jira) that are sophisticated *technologically* but naive *information-theoretically*—increasing bandwidth without understanding capacity, adding features without considering requisite variety, building interfaces without feedback loops.

Signal Theory fills both gaps: the disciplinary gap between the seven foundational theories, and the practical gap between information theory and the systems that implement it.

## 2.9 Additional Foundations

Seven additional works inform Signal Theory. Ackoff’s DIKW hierarchy (Ackoff, 1989) provides the Data → Information → Knowledge → Wisdom progression that maps to Signal lifecycle stages (§8.1), though it has been critiqued for oversimplification (Rowley, 2007; Frické, 2009). Zipf’s Principle of Least Effort (Zipf, 1949) provides the behavioral basis for the Path of Least Resistance (Section X). Conant and Ashby’s Good Regulator Theorem (Conant and Ashby, 1970)—“every good regulator of a system must be a model of that system”—establishes why the Optimal System must model its own communication environment. Maturana and Varela’s autopoiesis (Maturana and Varela, 1972) provides the foundation for self-maintaining systems (§8.7). Beer’s Team Syntegrity (Beer, 1994) provides the non-hierarchical coordination topology (§8.8). Kahneman’s dual-process theory (Kahneman, 2011) provides the cognitive basis for encoding optimization through pattern compilation (§10.6). Penrose’s non-computability thesis (Penrose, 1989) provides the principled boundary between computational Signal processing and human judgment (Section IX).

## 2.10 The Combined Constraint

The four governing principles serve a single root objective: **maximizing Signal-to-Noise Ratio**. S/N is not one metric among many—it is the fundamental measure of communication quality, the contrast between intent and noise. The four principles define the constraints on how that maximization is achieved:

1. **Shannon (the ceiling):** Channel capacity defines the maximum S/N achievable in any given channel—the system must encode below this limit
2. **Ashby (the repertoire):** Requisite variety ensures the system has enough Signal types to maintain S/N across every situation the environment produces
3. **Beer (the architecture):** The recursive five-system structure maintains S/N at every scale—from a single message to an entire organization
4. **Wiener (the correction):** Closed feedback loops detect when S/N is degrading and trigger corrective action

These four constraints constitute a comprehensive, non-redundant set—each addresses a distinct dimension of communication viability that the other three do not cover. The paper does not claim these four exhaust every possible constraint on communication systems; it claims that they are *individually necessary* for viable communication and

*collectively sufficient to diagnose the major failure modes* observed in practice. Whether additional constraints exist that produce failure modes not capturable by these four is an empirical question (see P5, Section 11.5). Subsequent sections apply, test, and extend them; they are not restated.

Any system missing any of these constraints is suboptimal. It will either exceed channel capacity (Shannon violation), fail to handle encountered situations (Ashby violation), lose viability when a subsystem is absent (Beer violation), or accumulate uncorrected errors (Wiener violation). The Optimal System satisfies all four constraints simultaneously, at every level of recursion.

The integration is not merely additive—it produces diagnostic power that no individual theory possesses. Consider a concrete case: a technical specification is sent to the correct team (routing is accurate), written in the correct genre (spec), with proper structure (requirements, acceptance criteria, dependencies)—yet the project fails. No single theory can diagnose this failure:

- **Shannon analysis** finds the channel was below capacity—not a bandwidth problem
- **Bakhtin analysis** confirms genre competence—the receiver can decode specs
- **Ashby analysis** confirms sufficient variety—the system has the right Signal types
- **Wiener analysis** reveals the failure: no feedback loop existed between the spec’s author and the implementers. The spec contained an ambiguous requirement that was decoded differently than intended, but the error was never detected because no mechanism existed for the implementers to confirm their interpretation before building. The error compounded through three sprints of development

The diagnosis requires all four theories simultaneously: Shannon confirms the channel is adequate, Bakhtin confirms the genre is correct, Ashby confirms the repertoire is sufficient—and only Wiener’s feedback principle identifies the actual failure mode. It takes the integrated framework to say: “This is a Feedback Failure (Section 6.4)—close the loop with a confirmation step after the spec is decoded.”

## 3 The Signal: Definition and Properties

### 3.1 The Problem of Naming

When an encoder must transmit a directive to a colleague, a data package to an analyst, a strategic vision to a team—what is the fundamental unit being created? What is the name for the encoded unit of intent that, regardless of form or channel, when decoded by the receiver, produces action?

No single field has a term that covers the full concept. Each names only its piece:

Field	Term	What It Covers	What It Misses
Information Theory (Shannon)	Signal	The encoded message traveling through the channel	Social conventions, structure, intent
Speech Act Theory (Austin/Searle)	Directive	A communication that compels the receiver to act	Mode, format, channel, structure
Semiotics (Kress, Chandler)	Text	Any coherent body of signs that can be interpreted	Only linguistic mode; misses intent
Business / Project Mgmt	Artifact	The tangible deliverable that carries the intent	Can be passive; misses directive function
Military / Creative Industries	Brief	A document containing everything needed to execute	Only one genre among many
Knowledge Management	Externalization	Encoding tacit knowledge into explicit transmittable form	Misses action, mode, convention

**Table 2:** Field-specific terminology for the fundamental unit of communication.

Each term captures one dimension of a multidimensional phenomenon. “Text” is mode-specific (linguistic only). “Document” is format-specific. “Artifact” is passive (it can sit on a shelf). “Directive” is type-specific (not all Signals are directives). None is simultaneously mode-independent, intent-carrying, channel-independent, and measurable.

### 3.2 Why Every Proposed Base Term Fails

The obvious candidate terms each fail for a specific reason:

Attempted Base Term	Why It Fails
Text	Mode-specific—a spoken Signal is not a “text,” a diagram is not a “text”
Document	Format-specific—a video is not a “document,” a conversation is not a “document”
Artifact	Passive—an artifact can sit on a shelf; a Signal demands action
Node	Topological—a node is where encoding/decoding happens, not the thing being encoded

**Table 3:** Why candidate base terms fail.

**Signal** succeeds because it is mode-independent, intent-carrying, channel-independent, measurable, and classifiable across five dimensions.

### 3.3 The Signal Defined (Resolution)

In this framework, a **Signal** is:

An encoded unit of intent that carries actionable information through a communication channel, designed to be decoded by the receiver into executable action. It is classified across five dimensions: Mode, Genre, Type, Format, and Structure.

A Signal is represented as a 5-tuple:

$$S = (M, G, T, F, W) \quad (2)$$

Where:

- $M \in \text{Modes} = \{\text{Linguistic, Visual, Auditory, Gestural, Spatial}\}$  (or combinations thereof)
- $G \in \text{Genres} =$  the set of conventionalized communicative forms recognized by the receiver
- $T \in \text{Types} = \{\text{Direct, Inform, Commit, Decide, Express}\}$  (or combinations thereof)
- $F \in \text{Formats} =$  the set of physical/digital containers available in the channel
- $W \in \text{Structures} =$  the set of internal structural templates associated with genre  $G$

A Signal is fully specified when all five dimensions are determined. The dimensions are not fully independent—they cascade. Format constrains Mode (an audio file cannot carry visual mode). Mode constrains Genre (a demo requires gestural mode; a report requires linguistic). Genre constrains Structure:  $W(G)$ , each genre carrying its own set of applicable structures, making the Signal space  $M \times G \times T \times F \times W(G)$  rather than a strict Cartesian product. **Type is the independent axis**—any type (Direct, Inform, Commit, Decide, Express) can ride any genre, mode, or format combination. The cascade is the encoder’s actual decision tree: what container  $\rightarrow$  what sensory channel  $\rightarrow$  what conventionalized form  $\rightarrow$  what internal skeleton, with what communicative function operating independently throughout.

The five dimensions are the encoder’s degrees of freedom. The **destination**—the receiver—is not a sixth dimension. It is the constraint field within which all five are optimized. You cannot choose the right genre without knowing what the receiver can decode, the right mode without knowing what they perceive, the right format without knowing their available channels. A Signal encoded without knowing its destination is a broadcast. Broadcasting is a Shannon violation—it ignores channel capacity by assuming all receivers are the same.

A Signal is **a change that enables action**. Gregory Bateson defined information as “a difference which makes a difference” (Bateson, 1972). A Signal is the *engineered* form of that difference—structured, intentional, and directed toward a specific change in the receiver’s capacity to act. Everything else is data waiting to become a Signal, or noise pretending to be one.



Decoding is not passive extraction—it is an interpretive act (Bakhtin, 1986; Hall, 1973). The same Signal may be decoded differently by different receivers depending on their position, context, and genre competence.

## 4 The Signal Taxonomy

A Signal has five distinct dimensions—each answers a different question about the Signal.

### 4.1 The Five Dimensions

```
SIGNAL (the encoded intent)
|
+-- MODE ----- How is it perceived? (sensory channel)
+-- GENRE ----- What conventionalized form does it take?
+-- TYPE ----- What does it DO? (communicative function)
+-- FORMAT ----- What is the container/vessel?
+-- STRUCTURE ---- How is it structured? (internal skeleton)
```

A sixth concept—the **Package**—denotes a bundle of Signals delivered together.

### 4.2 Dimension 1: MODE (How Is It Perceived?)

Mode is the sensory channel through which the Signal reaches the receiver (Kress and van Leeuwen, 2001).

Mode	What It Is	Examples
Linguistic	Written or spoken words	Document, email, voice memo, conversation
Visual	Images, diagrams, spatial arrangements	Miro board, flowchart, screenshot, video
Auditory	Sound, voice, music	Voice note, podcast, meeting recording
Gestural	Body language, demonstrations, walkthroughs	Live demo, screen share, in-person explanation
Spatial	Physical or virtual arrangement	Whiteboard layout, workspace organization, Figma canvas

**Table 4:** Signal modes and their characteristics.

A Signal can be **monomodal** (one mode) or **multimodal** (combining modes). A Loom video is multimodal—it combines linguistic (spoken words), visual (screen recording), and gestural (cursor movement) modes simultaneously.

Written text operates by the logic of *time* (sequential). Visual images operate by the logic of *space* (simultaneous arrangement). Choosing the wrong mode for a Signal is choosing the wrong logic for the message.

### 4.3 Dimension 2: GENRE (What Conventionalized Form Does It Take?)

Genre is the conventionalized, socially recognized form of communication (Bakhtin, 1986). A genre is not a format. A “proposal” is a genre—it has expected sections, conventions, and purposes that both sender and receiver understand. Whether that proposal is a PDF, a Google Doc, or spoken aloud in a meeting—it is still a proposal. The genre is the *pattern*. The format is the *container*.

Representative genres across organizational domains:

Genre	Domain	Purpose	Typical Receiver
SOP	Operational	Repeatable process execution	Operations, team members
Roadmap	Strategic	Strategic direction and sequencing	Product teams, stakeholders
Proposal	Strategic	Persuade toward a course of action	Decision-makers, clients
Brief	Strategic	Provide everything needed to execute	Content creators, designers
Spec	Technical	Define precise implementation requirements	Engineers, developers
ADR	Technical	Record a technical decision with context	Engineering teams
RFC	Technical	Propose a change and solicit feedback	Cross-functional engineers
Script (Media)	Creative	Direct content creation	Content creators, producers
Script (Software)	Creative	Automate a technical process	Developers, ops engineers
Contract	Legal	Bind parties to commitments	Legal, partners, clients
Report	Analytical	Inform on status or findings	Stakeholders, leadership
Post-Mortem	Technical	Analyze a failure to prevent recurrence	Engineering, leadership

**Table 5:** Representative genres across organizational domains.

The complete genre catalogue—all 36 genres with frameworks, templates, and use cases—is provided in the companion document *Signal Theory: Complete Genre Reference*.

A developer who writes excellent specs may struggle with proposals. A marketer who writes compelling briefs may be unable to produce a retrospective. This is not a knowledge gap—it is a genre competence gap (Bakhtin, 1986).

## 4.4 Dimension 3: TYPE (What Does It DO?)

Type is the communicative function of the Signal—what action it performs on the receiver (Searle, 1976). The labels are practitioner terms; Searle’s original categories are shown for academic reference.

Type	Searle Term	What It Does	Examples
Direct	Directive	Compels the receiver to act	SOP, task assignment, code review feedback, brief
Inform	Assertive	States facts or conveys information	Status update, report, data drop, dashboard
Commit	Commissive	Commits the sender to future action	Contract, promise, roadmap commitment, SLA
Decide	Declarative	Changes reality by being stated	Approval, termination, promotion, release announcement
Express	Expressive	Conveys attitude or evaluation	Praise, criticism, congratulations, retrospective reflection

**Table 6:** Signal types mapped to Searle’s taxonomy.

A single Signal can carry multiple types. A PR review is both Inform and Direct. The *intent* determines the type, not the words—“It’s cold in here” can be Inform, Direct, or Express depending on context.

## 4.5 Dimension 4: FORMAT (What Is the Container?)

Format is the physical or digital vessel that carries the Signal—strictly the **container**.

Format	What It Is	Modes It Can Carry
Document (PDF, DOCX, MD, Google Doc)	Written file	Linguistic, Visual
Slide Deck (Keynote, Google Slides, PPT)	Presentation file	Linguistic, Visual, Spatial
Spreadsheet (Excel, Google Sheets, CSV)	Tabular data file	Linguistic, Visual (charts)
Video File (MP4, Loom, recording)	Audio-visual recording	Linguistic, Visual, Auditory, Gestural
Audio File (MP3, voice memo, podcast)	Sound recording	Linguistic, Auditory
Interactive Canvas (Miro, Figma, FigJam)	Collaborative visual space	Visual, Spatial, Linguistic
Message (Email, Slack, SMS, DM)	Text-based communication	Linguistic
Live Session (Meeting, call, demo)	Synchronous real-time	All modes simultaneously
Codebase (Repository, PR, branch)	Source code and metadata	Linguistic (code as language)

**Table 7:** Signal formats and supported modes.

The same genre can exist in different formats. A roadmap can be a Google Doc, a slide deck, a Miro board, or a spoken presentation. The genre (roadmap) stays the same. The format (container) changes.

## 4.6 Dimension 5: STRUCTURE (How Is It Structured?)

Structure is the internal skeleton—the template or schema that organizes content *within* a genre. The genre is “proposal.” The structure is: Problem → Solution → Scope → Timeline → Cost → Terms. Without structure, the encoder transmits noise, not Signal.

Representative structures across domains:

Genre	Structure (Internal Skeleton)
SOP	Purpose → Scope → Responsibilities → Steps → Exceptions → Revision History
Proposal Brief	Problem → Solution → Scope → Approach → Timeline → Budget → Terms Objective → Background → Target Audience → Scope → Deliverables → Constraints → Timeline
Spec	Overview → Requirements → Acceptance Criteria → Constraints → Dependencies → Out of Scope
ADR	Title → Status → Context → Decision → Consequences → Alternatives Considered
Post-Mortem	Summary → Timeline → Impact → Root Cause → Contributing Factors → Corrective Actions
Script (Media)	Hook → Context → Body → CTA → Closing
Contract	Parties → Definitions → Obligations → Terms → Payment → Termination → Signatures

**Table 8:** Representative genre structures.

The complete structure catalogue for all 36 genres—including Runbooks, Roadmaps, RFCs, API Docs, Test Plans, Deployment Notes, Brand Guidelines, Data Drops, SLAs, SOWs, and others—is provided in the companion documents *Signal Theory: Complete Genre Reference* and *Signal Theory: Complete Taxonomy Reference*.

Structures are reusable. Once established for a genre, every subsequent Signal follows the same skeleton, reducing decoding effort—Shannon’s redundancy principle in action.

## 4.7 The Package

When multiple Signals must be delivered together—a collection of documents, assets, references, and directives—the result is a **Package**.

Package Type	What It Contains	Use Case
Dossier	Collection of documents on a single subject	Client research, competitive analysis
Brief Package	Brief + supporting materials + references	Agency handoff, creative project kickoff
Deliverable Package	Finished outputs bundled for handoff	Sprint delivery, milestone completion
Onboarding Kit	Guides + credentials + context + contacts	New hire orientation, new client setup
Handoff Package	Current state + history + next steps	Team transition, project transfer
Data Drop (as package)	Raw data + context doc + action items	Content creator info dump, research handoff
RFP / Bid Package	Requirements + criteria + submission instructions	Vendor selection, procurement

**Table 9:** Package types and use cases.

A Package is not a single Signal—it is a Signal Network pattern (Section VI) where multiple Signals converge on the same destination for the same purpose. Each Signal in the Package may use a different mode, genre, or format, but all are aligned to the same receiver and the same intended action—different vectors, same target.

## 5 Signal Encoding: Principles and Constraints

The quality of communication is determined at the point of encoding.

### Encoding Principles

Six principles govern encoding quality:

1. **Mode-message alignment**—Sequential logic (processes, steps) maps to linguistic mode. Relational logic (systems, dependencies) maps to visual mode. Emotional or nuanced content maps to auditory/gestural modes. Selecting the wrong mode is selecting the wrong logic for the message.
2. **Genre-receiver alignment**—A developer decodes a spec efficiently. A content creator decodes a brief. A client decodes a report. Transmitting the wrong genre to the wrong receiver constitutes encoding failure—not because the information is wrong, but because the receiver lacks genre competence for that form.
3. **Structure imposition**—Raw information is noise. Structure converts noise into Signal. Every genre has an optimal internal skeleton; when none exists, establishing one and making it repeatable converts future encoding from a deliberative to a reflexive process.
4. **Redundancy proportional to noise**—Shannon proved that redundancy enables error correction. Noisy channels (asynchronous, cross-timezone, complex subject

matter) require proportionally greater structure, context, and explicit intent.

5. **Entropy preservation**—Entropy is information content. Every Signal should carry maximum meaning with minimum waste. An SOP that buries the action in paragraphs of context has lost entropy to noise.
6. **Bandwidth matching**—A 50-page spec sent to a receiver who needs 3 bullet points exceeds the receiver’s channel capacity. Three bullet points sent when the task requires a spec falls below the channel’s information requirement. The Signal must match the receiver’s decoding bandwidth.

## 6 The Signal Network

In Shannon’s model, every communication requires a source and a destination. In an organization, those sources and destinations form a network. Signal Theory adds a precise description of the **function** each entity serves in that network—the organizational name stays, but the signal function enables diagnosis and optimization.

### 6.1 Network Architecture

Shannon’s Model	-> Organizational Reality
Information Source	-> The person/team encoding intent (CEO, PM, Lead)
Transmitter (Encoder)	-> The act of creating the Signal (writing, recording, designing)
Channel	-> The medium (email, Slack, meeting, doc share)
Noise	-> Organizational interference (politics, overload, ambiguity)
Receiver (Decoder)	-> The person/team interpreting the Signal
Destination	-> The action taken at the endpoint

### 6.2 Organizational Entities and Their Signal Network Functions

Every organizational entity has a dual identity: what it **is** (its organizational name) and what it **does** in the signal network (its network function). Both names are valid. The organizational name describes its purpose. The network function describes how it participates in communication.

Organizational Entity	Signal Network Function	Description
Company / Organization	Network	The bounded system within which Signals flow
Team	Cluster	Tightly-connected endpoints with shared genre competence
Person / Individual	Endpoint	A source or destination that encodes or decodes Signals
Project	Signal Chain	A sequence of signals directed toward a bounded outcome
Partner / External Entity	Bridge	A connection between two separate Networks
Machine Intelligence	Signal Processor	Non-human nodes that encode, decode, translate, and route Signals across genres and structures

**Table 10:** Organizational entities and their signal network functions.

The complete entity mapping—including Departments (Subnets), Recurring Processes (Standing Channels), Products (Signal Targets), and Knowledge Bases (Signal Memory)—is provided in the companion document *The Optimal System: Architecture Specification*.

### 6.3 Signal Routing

The relationships between organizational entities describe how Signals move through the network:

Relationship Type	Signal Theory Equivalent	Example
Hierarchy (reporting structure)	Signal flows through authority gradient—each level re-encodes for the next	CEO → VP → Manager → IC: the same strategic intent is re-encoded at each level to match the receiver's bandwidth
Dependency	Signal B cannot be encoded until Signal A has been decoded and acted upon	The deployment cannot happen until the code review Signal has been decoded and approved
Reference	One Signal references another as context	A PR description references the ADR that explains the architectural decision
Collaboration	Two-way Signal flow between entities	Frontend and backend teams exchanging API contract signals
Sequence	Ordered signal chain—Signal 1 must be decoded before Signal 2 is sent	Brief → Draft → Review → Revision → Approval → Publish

**Table 11:** Signal routing relationships in the network.

## 6.4 Failure Modes in the Network

Each failure mode maps to a governing constraint violation: **Shannon** (Routing, Bandwidth Overload, Fidelity), **Ashby** (Genre Mismatch, Variety, Structure), **Beer** (Bridge, Herniation, Decay), **Wiener** (Feedback). Adversarial Noise operates across all four. Eleven modes, four groups—the diagnostic question is always: *which constraint is violated?*

<b>Constraint</b>	<b>Failure</b>	<b>Diagnostic Question</b>	<b>Corrective Action</b>
Shannon	Routing Failure	Did the right person/team receive this Signal?	Map the Signal chain; re-route to correct endpoint
	Bandwidth Overload	Is the receiver overwhelmed with incoming Signals?	Reduce volume, prioritize, batch, or delegate
	Fidelity Failure	Did the receiver extract the intended meaning?	Re-encode with clearer structure, more redundancy, or different mode
Ashby	Genre Mismatch	Does the receiver know how to decode this type of Signal?	Re-encode in a genre the receiver recognizes, or train genre competence
	Variety Failure	Does the system have a genre/mode/structure for this situation?	Expand the repertoire—create new genres, structures, or modes
	Structure Failure	Is the Signal organized in a way the receiver can parse?	Apply or create a structure for the genre; impose skeleton before sending
Beer	Bridge Failure	Do both sides share enough context to decode each other?	Establish shared conventions; add context preambles; assign a bridge node
	Herniation Failure	Does this Signal maintain coherence across all semantic layers?	Map against all affected layers; re-encode with proper layer traversal
	Decay Failure	Is this Signal still accurate, or has the environment changed?	Audit stored Signals on a cadence; version them; sunset outdated ones
Wiener	Feedback Failure	Does the sender know whether the receiver decoded correctly?	Close the loop—add confirmation steps, check-ins, or retrospectives
All	Adversarial Noise	Is someone deliberately degrading Signal quality?	Identify the actor; create redundant channels; escalate; make noise visible

**Table 12:** Eleven failure modes mapped to governing constraint violations.

The failure modes are diagnostic categories, not mutually exclusive—a single failure event frequently involves multiple modes simultaneously, as the Appendix D walkthrough demonstrates (triple failure: Variety + Architecture + Feedback).

#### 6.4.1 Error Severity Classification

Not all Signal errors are equal. A failure taxonomy requires severity classification to prevent over-rejection of Signals with minor flaws and under-rejection of Signals with structural defects.

**Critical errors** invalidate the Signal: logical contradiction, fundamental misunderstanding of the receiver’s domain, invalid causal inference, or herniation across architectural layers. These require re-encoding.

**Non-critical errors** do not invalidate sound reasoning: terminology imprecision, formatting inconsistency, stylistic deviation from genre convention. A Signal with 80%+ semantic correctness and minor surface errors should not be rejected—this mistakes syntactic noise for semantic failure. The corrective action is annotation, not re-encoding. Shannon’s noisy channel theorem applies: the intended message can be recovered despite surface-level noise, provided the underlying structure is intact.

Each failure type has a different fix. The diagnostic framework above enables systematic identification of failure type, which determines the corrective action.

## 7 Signal Composition: The Micro-Structure

Every Signal is built from smaller units—“atoms” that combine into larger structures. The linguistic compositional hierarchy (grapheme → morpheme → word → sentence → paragraph → section → work) is well established, but it describes the structure of only **one mode: Linguistic**. Every mode has its own compositional hierarchy—Visual (mark → shape → element → composition → canvas), Auditory (phoneme → utterance → turn → conversation), Code (token → expression → function → module → codebase).

The Structure dimension (Dimension 5) defines the **macro-structure** of a Signal—the skeleton. The compositional hierarchy defines the **micro-structure**—what the skeleton is built from. When you write an SOP (Genre), its Structure is Purpose → Scope → Responsibilities → Steps → Exceptions. But each “Step” is composed of sentences (Linguistic mode), which are composed of words, which are composed of morphemes. The Structure tells you WHAT sections exist. The compositional hierarchy tells you HOW each section is built at the atomic level. The complete compositional hierarchies for all modes, with examples at each level, are provided in the companion document *Signal Theory: Signal Composition*.

The distinction matters for diagnosis. A spec with the correct Structure (Overview → Requirements → Acceptance Criteria → Dependencies) but poorly constructed sentences within each section is a **composition failure**—the micro-structure is degraded while the macro-structure is intact. A document with well-crafted prose but no recognizable section organization is a **Structure failure**—the macro-structure is missing while the micro-structure is fine. The corrective actions are distinct: composition failures require

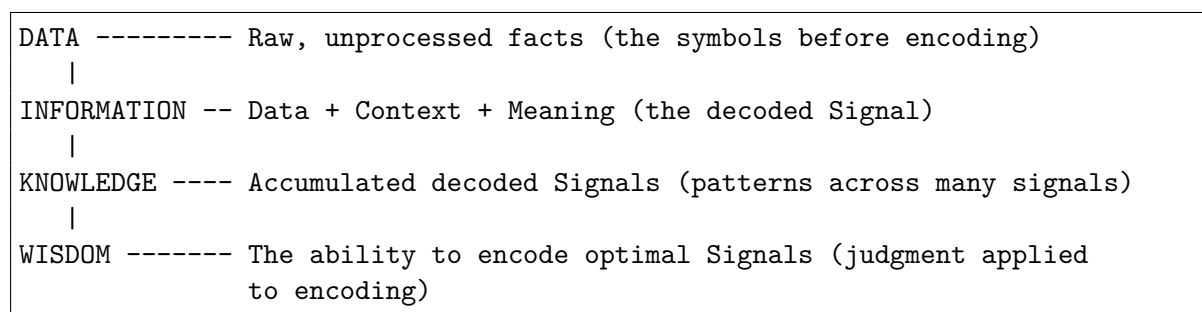
re-writing at the sentence and paragraph level; Structure failures require imposing the genre skeleton. The framework distinguishes these because the fix is different.

## 8 The Optimal System Architecture

Section I posed a question: *How does one encode a message in a multidimensional environment where things are constantly changing but are never changed by them?* The preceding sections define the components of the answer—the Signal (the payload), the Taxonomy (the classification), the Encoding Principles (the quality constraints), the Network (the topology), and the Composition (the micro-structure). This section assembles them into the complete Optimal System architecture.

### 8.1 Data, Information, and the Signal Lifecycle

Data and Information are not the same thing. In Signal Theory terms:



This maps to the Signal lifecycle: Data is captured, encoded into a Signal (becoming Information), accumulated across many decoded Signals (becoming Knowledge), and eventually internalized as the judgment that produces better encoding (Wisdom). The DIKW hierarchy ([Ackoff, 1989](#)) provides the foundational progression, though Signal Theory focuses on the transitions—the encoding and decoding acts—rather than the static levels.

### 8.2 The Interface Layer

The Optimal System has **Interfaces**—the display and interaction layer where decoded Signals (information) are presented to human endpoints. An Interface is a **decoding surface**—it transforms raw data or stored Signals into human-perceivable, actionable information. Dashboards, project management tools, communication platforms, CRMs, code environments, and knowledge bases are all interfaces—each specialized for the Signal types its domain produces and consumes. An Interface does not create Signals—it decodes and displays them. The Interface IS the decoder in Shannon’s model, implemented as software.

Standard interface categories have emerged for each business function through convergent evolution—CRMs, ERPs, ticketing systems, project management tools—because the Signal processing requirements of each function dictate the optimal interface design. This is Ashby’s Law applied to interface design: each function’s interface must have sufficient variety to handle the Signal types that function produces and consumes. When it does not,

the organization compensates with workarounds (spreadsheets, Slack messages, manual tracking)—symptoms of a variety failure at the interface layer.

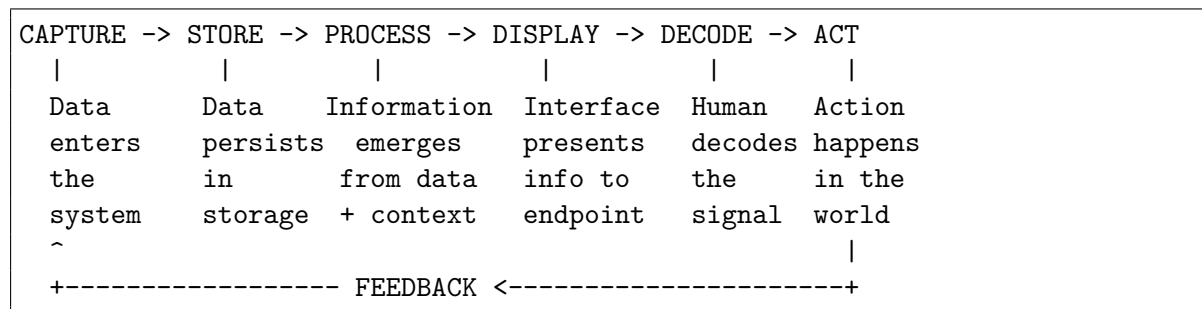
### 8.2.1 The Silo Constraint and Its Resolution

These standard interfaces were individually optimal but collectively siloed. A CRM cannot natively decode Signals from an ERP. Inter-system communication required human relay—re-encoding at each handoff, with encoding loss at each step.

Machine Intelligence has begun dissolving this constraint. AI agents bridge siloed interfaces—reading from one system, composing data from another, and producing a Signal in the genre the receiver expects. The Optimal System approaches fuller achievability as Machine Intelligence enables the governing principles to be satisfied across the entire Signal network, not just within individual interfaces.

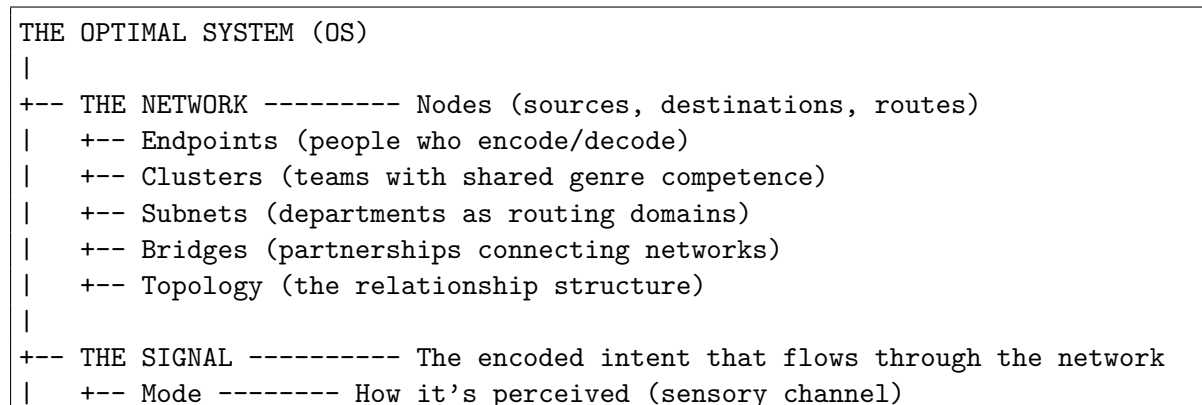
Every major communication technology has changed the Interface layer while leaving Shannon’s architecture intact—from written text (~3200 BCE) through the printing press, telegraph, telephone, GUI, web, smartphones, and modern SaaS tools to emerging AI-assisted systems. The decoding surface evolves; the underlying model (Source → Encoder → Channel → Decoder → Destination) does not. Signal Theory classifies at the level of communicative function precisely because this level persists across every technological shift. The full historical progression is traced in Appendix C.

## 8.3 How Data Moves Through the System



The results of action are captured as new data—the loop closes (Wiener). This is Shannon’s linear model made cybernetic.

## 8.4 The Complete Architecture



```

| +-- Genre ----- What form it takes (conventionalized pattern)
| +-- Type ----- What it does (communicative function)
| +-- Format ----- What container it's in (physical/digital vessel)
| +-- Structure --- How it's internally structured (skeleton)
| +-- Package ----- Bundles of Signals delivered together
|
+-- THE COMPOSITION ----- The micro-structure of each Signal
| +-- Linguistic (grapheme -> word -> sentence -> section -> work)
| +-- Visual (mark -> shape -> element -> composition -> canvas)
| +-- Auditory (phoneme -> utterance -> turn -> conversation)
| +-- Code (token -> expression -> function -> module -> codebase)
|
+-- THE INTERFACE ----- The display/interaction layer
| +-- Dashboards (data -> visual information)
| +-- Project Tools (signal chains -> sequenced tasks)
| +-- Communication Platforms (real-time signal exchange)
| +-- Knowledge Bases (accumulated signals -> searchable knowledge)
| +-- Custom Interfaces (domain-specific decoders)
|
+-- THE DATA LAYER ----- Storage, movement, and transformation
| +-- Capture (data enters the system)
| +-- Store (data persists)
| +-- Process (data + context = information)
| +-- Display (information presented through interface)
| +-- Archive (signals preserved for future reference)
|
+-- THE FEEDBACK LOOP --- Circular causality and self-correction
| +-- Output -> Measurement -> Comparison -> Error Signal -> Correction
| +-- Bidirectional flow at every node
| +-- Enables iterative improvement of Signal fidelity
|
+-- THE VIABLE STRUCTURE Beer's five recursive subsystems
| +-- System 1: Operations (produce/consume Signals)
| +-- System 2: Coordination (prevent Signal conflicts)
| +-- System 3: Control (manage Signal flow and resources)
| +-- System 4: Intelligence (scan for external Signals)
| +-- System 5: Policy (define encoding standards and identity)
|
+-- THE PRINCIPLES ----- The four governing constraints (Section 2.10)

```

Any system implementing all components, satisfying all four principles, at every level of recursion, is an Optimal System.

## 8.5 Implementation Evidence: Erlang

The architecture described above is not theoretical. The Erlang programming language ([Armstrong, 2003](#)), created at Ericsson in 1986, encoded Shannon's principles directly into a system architecture. Named after Agner Krarup Erlang—the mathematician who founded teletraffic engineering—Erlang implemented each principle as a language primitive:

Shannon Principle	Erlang Implementation
Message passing through channels	Isolated processes communicate only via messages—no shared state
Noise is inevitable	“Let it crash”—design for failure recovery, not failure prevention
Redundancy enables reliability	Supervision trees—redundant observer processes monitor and restart workers
Channel capacity is finite	Each process has its own bounded mailbox (backpressure when full)
Encoding can be changed without disrupting the channel	Hot code swapping—update the encoder while the system runs

**Table 13:** Shannon’s principles implemented in Erlang.

The systems built on Erlang and its principles—WhatsApp (billions of messages daily), global telecom switching infrastructure, internet routing (Cisco)—achieve extraordinary reliability not by preventing failure, but by implementing Shannon’s architecture: bounded channels, redundant encoding, feedback loops, and graceful degradation. When the Optimal System’s architecture is implemented as system design rather than merely applied as theory, it produces communication systems of exceptional reliability and scale.

## 8.6 Implementation Evidence: The Toyota Production System

Where Erlang demonstrates Shannon’s principles implemented as software, the Toyota Production System ([Ohno, 1988](#)) demonstrates all four governing principles implemented as organizational practice—discovered independently through decades of manufacturing iteration:

Governing Principle	Toyota Implementation
Shannon	Kanban cards carry exactly the information needed—no more. Visual management boards keep Signal volume within human processing bandwidth. One card per bin; information density matches channel capacity.
Ashby	Standardized Work converts novel encoding (System 2) into repeatable encoding (System 1). The system has specific Signal types for every recurring situation: kanban for pull, andon for escalation, A3 for problem-solving, hoshin kanri for strategic alignment.
Beer	Each work cell is a viable system (operations, coordination, control, intelligence, policy). Each line is a viable system of cells. Each plant is a viable system of lines. The five-system structure is recursive at every scale.
Wiener	The andon cord—any worker who detects a defect stops the line immediately, correcting the error at the source rather than letting it compound downstream. Daily stand-ups, visual boards, and the A3 process close feedback loops at every level.

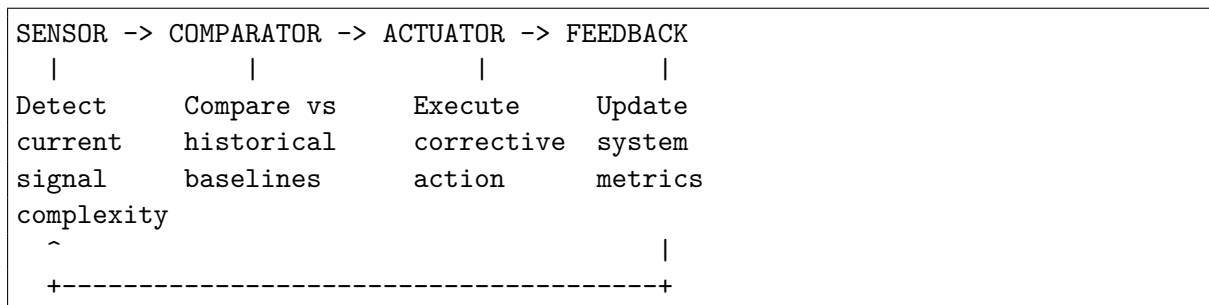
**Table 14:** The four governing principles in the Toyota Production System.

Toyota’s core metric—maximizing value while eliminating *muda* (waste)—is the manufacturing equivalent of Signal-to-Noise Ratio. The convergence is not coincidental: any system designed to maximize useful output in a noisy, failure-prone environment, whether software or factory floor, discovers the same governing constraints.

## 8.7 Autopoiesis: The Self-Maintaining System

The Optimal System is not merely well-designed—it is **self-maintaining**. Beer drew on Maturana and Varela’s concept of autopoiesis (Maturana and Varela, 1972): a system that continuously regenerates itself from within, maintaining its own organization despite environmental perturbation.

In Signal Theory terms, an autopoietic communication system monitors its own Signal complexity, detects degradation trajectories, and restructures before critical failure. The mechanism is a cybernetic homeostat:



The critical insight is that autopoietic systems track **trajectories**, not just snapshots. A genre that grows more complex with each iteration, a channel whose noise floor is rising over time, a feedback loop whose latency is increasing—these are **tumor signals**, components whose complexity is growing in a direction that will eventually herniate across architectural boundaries. The autopoietic system detects these trends and restructures

proactively—simplifying the genre, rerouting the channel, tightening the feedback loop—before the failure mode manifests.

This extends Beer’s Viable System Model: System 4 (Intelligence) is the subsystem that scans the external environment, but autopoiesis adds **self-scanning**—the system monitors its own internal communication architecture with the same rigor it monitors external perturbations. An organization that reviews its own communication patterns—auditing Signal quality, measuring re-encoding frequency, detecting genre drift—is implementing autopoietic maintenance. An organization that does not is accumulating structural debt that will manifest as cascading communication failures.

## 8.8 Syntegrity: Non-Hierarchical Coordination

Beer’s work extended beyond the VSM into an alternative coordination topology: **Syntegrity** (syntactic integration). Based on the icosahedron—a 20-faced polyhedron—Syntegrity structures group decision-making with zero hierarchy:

- **30 participants** (one per edge)—each connects to exactly 2 topics
- **12 topics** (one per vertex)—each has exactly 5 participants
- **20 working groups** (one per face)—each addresses a unique intersection of topics
- **Perfect symmetry**—no participant has structural advantage over any other

The result is that 12 topics are processed simultaneously across 20 working groups, with each participant contributing to exactly 2 topics. Traditional hierarchical coordination would process these topics sequentially—12 meetings, each with a subset of relevant voices. Syntegrity processes them in parallel with full cross-pollination, eliminating approximately two-thirds of the coordination overhead.

This is relevant to Signal Theory because it demonstrates that **hierarchy is not the only viable coordination topology for Signal routing**. The VSM describes the recursive hierarchical structure; Syntegrity describes the non-hierarchical alternative. Both satisfy the governing principles—but through different topological configurations. The Optimal System may employ both: hierarchical routing for operational Signal flow (System 1–3 in the VSM) and syntegritic coordination for strategic multi-topic decision-making (System 4–5).

The architecture above—data layer, interfaces, feedback loops, viable structure, autopoietic maintenance, and both hierarchical and non-hierarchical coordination—was designed for human participants. The next section addresses what changes when the participants include machines.

# 9 Machine Intelligence in the Signal Network

## 9.1 The Signal Processor

The architecture described in this paper has historically assumed human endpoints—people who encode and decode Signals. That assumption no longer holds. **Machine Intelligence**—AI agents, machine learning models, algorithms, real-time data processors, and automated pipelines—is an operational class of node in the Signal network:

non-human entities that encode, decode, translate, route, aggregate, and compose Signals across genres, modes, and frameworks. Agent teams now work alongside human teams, handling the encoding and decoding load that previously consumed human cognitive bandwidth.

Machine Intelligence encompasses multiple forms of Signal processing: algorithms (deterministic transformation), machine learning models (pattern recognition), real-time processors (streaming transformation), data pipelines (multi-step aggregation), and AI agents (full autonomous Signal participants). These forms combine in practice—the entire chain operates as a unified **Signal Processor**. The model architectures map directly to Signal Theory’s Mode dimension: language models process linguistic Signals, vision models process visual Signals, speech models process auditory Signals, multimodal models translate between modes. The taxonomy is not tied to any specific model architecture—as capabilities evolve, the Signal functions remain constant.

This operates at every level of granularity—from strategic board reports down to a team member synthesizing priorities from multiple data sources into a single decision about what to work on next. The same architecture applies at every level, with Machine Intelligence performing the encoding and the interface performing the decoding.

## 9.2 The Computational Boundary

**The boundary between human and machine is operational, not categorical.** Roger Penrose argued (Penrose, 1989) that certain aspects of human understanding involve non-computable processes—grounded in Gödel’s incompleteness theorems. Whether or not one accepts the full Penrose thesis, Signal Theory adopts the operational distinction: the **computable** aspects of communication—pattern matching, classification, translation, routing, structural verification, genre compliance checking—are the aspects Machine Intelligence handles. The aspects that remain with human cognition fall into two categories: those that are **provably non-computable** in the Gödel-Turing sense and those that are **practically beyond current computation** (intent formation, judgment under ambiguity, meaning-making in novel contexts, ethical evaluation). The second category may narrow; the first cannot. Signal Theory requires only the operational principle: what can be algorithmic should be algorithmic, freeing human bandwidth for what cannot yet be, or can never be. By offloading the computable dimensions of Signal processing to Machine Intelligence, the effective processing bandwidth of the human endpoint increases.

## 9.3 Cross-Genre Encoding

**Every business function has its own Signal conventions**—Sales uses Prospect Brief → Proposal → Contract sequences, Engineering uses Spec → ADR → PR → Post-Mortem sequences—but the underlying structure is identical: intent is encoded, transmitted through a channel, and decoded into action. Machine Intelligence that can translate between these conventions is performing **cross-genre encoding**—the same function that human managers perform when they re-encode a strategic directive into department-specific Signals, but at organizational scale and without the lossy re-encoding that occurs at each human relay. The complete mapping of Signal chains across all business functions is provided in the companion document *The Optimal System: Architecture*

*Specification.*

**The architecture appears to be structural, not cultural.** When AI agent teams coordinate on complex tasks—building compilers, managing engineering organizations, triaging issues across repositories—they converge on hierarchy, task decomposition, dependency tracking, peer-to-peer messaging, and feedback loops. These are the same patterns described by Beer’s Viable System Model, the same structures that human organizations developed over centuries of trial and error. This convergence suggests that the governing principles are **structural constraints**, not cultural artifacts. If any intelligence—human or machine—coordinating communication at scale must satisfy the same requirements (bounded channels, sufficient variety, recursive architecture, closed feedback loops), then the patterns are not imposed by human convention but discovered under constraint. The extent to which Machine Intelligence independently converges on the same organizational patterns that humans evolved is an empirical question—but preliminary evidence from multi-agent coordination systems points toward structural necessity rather than cultural inheritance.

## 9.4 Governing Principle Satisfaction

**Machine Intelligence satisfies the governing constraints computationally**—monitoring channel capacity in real time, expanding its genre repertoire dynamically, implementing recursive coordination structures, and maintaining feedback loops without fatigue or attention decay. Effective Machine Intelligence systems work because they implement the architecture described in this paper. Those that fail violate the same constraints that cause human communication systems to fail.

# 10 The Path of Least Resistance

## 10.1 The Optimization Problem

The Introduction posed a question: *How does one encode a message in a multidimensional environment where things are constantly changing but are never changed by them?*

The environment is not a single channel with fixed noise—it is a multidimensional space where channel noise fluctuates, receiver bandwidth varies, genre competence differs across receivers, organizational context evolves, competing signals multiply, and environmental conditions shift. The encoder did not cause these changes and cannot control them. The question is not how to encode perfectly for a fixed environment—it is how to encode **robustly** across environmental states the encoder cannot predict or control.

## 10.2 The Encoder’s Degrees of Freedom

What the encoder controls is precisely what the five dimensions describe: **Mode** (sensory channel), **Genre** (conventionalized form), **Type** (communicative function), **Format** (container), and **Structure** (internal skeleton). These five choices are the encoder’s degrees of freedom. Everything else is environment.

### 10.3 Encoding Quality as Probabilistic Investment

Shannon’s error-correcting codes do not eliminate noise—they make the signal robust enough to survive the noise it encounters. Encoding quality is therefore a **probabilistic investment**. A well-encoded Signal does not guarantee successful transmission, but it succeeds across a wider range of environmental states. Over many Signals, the system that encodes well outperforms the system that encodes poorly, because its Signals succeed in conditions where the other system’s fail. Signal quality shows itself in outcomes over time—projects completed as specified, teams aligned, systems built to spec.

### 10.4 The Path Defined

Shannon’s channel coding theorem proves that optimal encoding schemes exist. Zipf’s Principle of Least Effort (Zipf, 1949) establishes that systems naturally evolve toward minimum-effort communication patterns. Morse code demonstrated this empirically a century before Shannon—channel pressure naturally selects for efficient encoding.

The path of least resistance is where these converge. It is not a metaphor. It is the encoding configuration that achieves maximum fidelity with minimum total system effort. What the receiver experiences as clarity, elegance, or beauty is the perceptual signature of optimal compression—maximum meaning per unit of signal, minimum noise per unit of meaning. Poetry compresses emotion. Elegant code compresses logic. A clean brief compresses intent. The path of least resistance is what optimal encoding *feels like* from the decoder’s side.

In applied terms, the path of least resistance is the Signal that:

- Uses the **mode** the receiver perceives most naturally
- Takes the **genre** the receiver has competence to decode
- Performs the **type** (speech act) that matches the intended outcome
- Arrives in the **format** appropriate to the channel
- Is organized by a **structure** that makes the internal skeleton decodable
- Carries enough **redundancy** to survive the noise it will encounter
- Minimizes **decoding effort** at the receiver
- Produces **action** with the highest probability across environmental states

### 10.5 The Path Before the Theory: Morse Code

The most striking evidence that the path of least resistance is a natural attractor—not a theoretical ideal—is that it was discovered empirically a full century before Shannon formalized it. Samuel Morse and Alfred Vail, developing the telegraph code between 1837 and 1844, faced a pure channel optimization problem: transmit messages through a noisy electrical channel with maximum throughput and minimum error.

Vail’s solution was to count letter frequencies by examining a typesetter’s letter cases at the Morristown newspaper, then assign code lengths inversely proportional to frequency:

Letter	Frequency Rank	Morse Code	Code Length
E	1st (most common)	.	1 symbol
T	2nd	-	1 symbol
A	3rd	.-	2 symbols
I	4th	..	2 symbols
Q	25th	-.-	4 symbols
J	26th (least common)	.-.-	4 symbols

**Table 15:** Morse code: variable-length entropy encoding.

Morse code is a **variable-length entropy code**—common letters get shorter signals, rare ones get longer, minimizing average signal length. This is the same principle Shannon proved mathematically in 1948, and that Huffman formalized as an optimal algorithm in 1952 ([Huffman, 1952](#)):

Year	Contribution	What It Established
1844	Morse/Vail (Morse Code)	Empirical variable-length encoding—frequency-based code assignment discovered by counting letters in a print shop
1948	Shannon (Information Theory)	Mathematical proof that variable-length codes can approach channel capacity
1952	Huffman (Huffman Coding)	Optimal algorithm for constructing minimum-redundancy codes

**Table 16:** The progression from empirical to formal encoding theory.

Each step formalized what the previous step discovered. Morse code is to Shannon’s theorem what Kepler’s observations were to Newton’s laws—the empirical evidence that preceded the theory, already embodying the principles the theory would later explain. Shannon himself cited Morse code in his 1948 paper as an example of a code that approximates optimal encoding.

The fact that an optimal encoding scheme was discovered by a machinist counting letters in a print shop—before the theory existed—is the strongest possible evidence. The path of least resistance is not imposed by theory. It is discovered under constraint.

## 10.6 Dual-Process Encoding

The path of least resistance has two modes, corresponding to Kahneman’s dual-process theory of cognition ([Kahneman, 2011](#)):

**System 2 (deliberative):** Novel Signals require full encoding effort—the encoder must consciously choose mode, genre, type, format, and framework. This is expensive. It consumes attention bandwidth. It is necessary when the Signal has no precedent—the first time a team writes an architectural decision record, the first time a sales team creates a proposal for a new market segment, the first time a cross-functional initiative requires a coordination Signal that no existing genre covers.

**System 1 (reflexive):** Recognized Signals trigger compiled responses—the encoder reaches for the established genre, applies the known structure, transmits through the familiar channel. This is near-zero cognitive load. The genre does the encoding work. The human applies judgment to content, not structure.

**The Optimal System learns to convert System 2 signals into System 1 patterns.** Every novel Signal that succeeds becomes a genre template. Every genre template that proves reliable becomes a standing pattern. The path of least resistance gets shorter through use—not because the channel changes, but because the encoding effort decreases as patterns compile from deliberative to reflexive. This is why genre establishment is so critical: a team that creates and adopts a genre for a recurring communication need converts an expensive System 2 process into a near-free System 1 process for every subsequent instance.

This also explains why **fewer, more fundamental operations outperform a larger library of specialized techniques.** An encoder with 500 formatting templates must search the template space before encoding—the search cost itself consumes bandwidth. An encoder with 5 core operations (RECOGNIZE the receiver’s needs, DECOMPOSE the intent into components, STRUCTURE using the appropriate framework, VERIFY against the genre convention, TRANSMIT through the optimal channel) handles more variety with less effort, because the operations work at the semantic level rather than the syntactic level. Quality of encoding operations matters more than quantity.

## 10.7 Signal Topology: Sequential and Parallel

Signals flow through the network in two fundamental temporal topologies:

**Sequential (linear):** Each Signal must be decoded before the next is sent. Signal B depends on Signal A’s decoded output. This is the default mode of hierarchical communication—CEO encodes → VP decodes and re-encodes → Manager decodes and re-encodes → Individual Contributor decodes and acts. Each step blocks until the previous completes.

**Parallel (concurrent):** Multiple Signals are transmitted simultaneously through independent channels. No Signal depends on another’s decoded output. This is the mode of cross-functional coordination—marketing, engineering, and product each receive their domain-specific Signals concurrently, decode independently, and converge at integration points.

The Optimal System uses both topologies strategically. Sequential flow is necessary where Signals have **causal dependencies**—the deployment Signal cannot be sent before the code review Signal is decoded and approved. Parallel flow is necessary where Signals are **causally independent**—the marketing campaign brief and the engineering sprint spec can be encoded and transmitted simultaneously because neither depends on the other’s decoded output.

The pathology is forcing sequential flow where parallel flow is possible. An organization that routes every Signal through a single decision-maker (sequential bottleneck) is wasting the parallel bandwidth available in its network. The Signal backs up, cognitive load concentrates at the bottleneck, and the system’s effective throughput drops below its theoretical channel capacity. This is a Shannon violation created by topology choice, not by encoding quality.

## 10.8 The Optimal System as Adaptive Architecture

The Optimal System is not a static configuration. It is a **continuously adaptive architecture** that maintains maximum Signal-to-Noise Ratio across all dimensions as those dimensions shift. The governing constraints converge on this adaptability: the channel ceiling moves as noise changes, the required variety evolves with the environment, System 4 detects change before failure, and the feedback loops detect when re-encoding is needed. The path of least resistance is not a fixed route—it is the route that the system discovers and rediscovers as the landscape shifts beneath it.

## 11 Discussion

### 11.1 Implications

Signal Theory has four primary implications:

**Measurability.** The question “is this a good document?” becomes “does this Signal encode the intended action in the mode, genre, type, format, and structure that the receiver can decode with maximum fidelity?”

**Diagnosis.** Communication failures can be systematically classified—routing failure, genre mismatch, structure failure, adversarial noise—and each type has a different fix.

**Design.** Rather than asking “how should we communicate?”—an open-ended question—the framework asks: “Does our system satisfy Shannon, Ashby, Beer, and Wiener at every level?” This is a bounded, answerable question.

**Recursion.** The architecture applies at every scale—from a single team to an entire organization. The same diagnostic framework works at every level.

### 11.2 Relationship to Existing Frameworks

Signal Theory is not the first attempt to model organizational communication, and its relationship to existing frameworks must be stated precisely—both to acknowledge intellectual debts and to clarify what, if anything, this integration adds.

**Luhmann’s Social Systems Theory** (Luhmann, 1984) is the most rigorous competitor. Luhmann models communication as a three-part selection: (1) selection of *information* (what is communicated), (2) selection of *utterance* (how it is communicated), and (3) selection of *understanding* (how the receiver interprets it). This maps partially to Signal Theory’s encoding/transmission/decoding architecture, but with a critical difference: Luhmann’s third selection—understanding—is constitutive of communication itself, not merely a decoding step. For Luhmann, communication does not occur until the receiver selects an understanding; for Signal Theory, the Signal exists once encoded and transmitted, and understanding is measured through the feedback loop. The difference is ontological: Luhmann defines communication as a social system operation that includes reception; Signal Theory defines the Signal as an engineering artifact whose reception is verified through feedback. Signal Theory does not subsume Luhmann—it operates at a different level of analysis. Where Luhmann asks “what constitutes communication as a social operation?”, Signal Theory asks “what architectural constraints must hold for

communication to achieve its intended effect?” The questions are complementary, not competing.

**Weick’s Sensemaking** (Weick, 1995) describes how receivers decode ambiguous Signals—a process Signal Theory models as the interpretive dimension of decoding (Section 3.6). Weick’s contribution is the phenomenology of the decoder; Signal Theory’s contribution is the architecture of the system within which decoding occurs. The frameworks are compatible: sensemaking is what happens inside the decoder, and Signal Theory provides the system-level constraints that determine whether sensemaking succeeds or fails.

**Nonaka’s SECI Model** (Nonaka and Takeuchi, 1995) maps to Signal Theory’s encoding/decoding cycle: Socialization (tacit-to-tacit) corresponds to gestural/auditory Signal exchange, Externalization (tacit-to-explicit) corresponds to encoding, Combination (explicit-to-explicit) corresponds to cross-genre translation, and Internalization (explicit-to-tacit) corresponds to the receiver’s conversion of decoded Signals into knowledge. Signal Theory adds the channel constraints and failure modes that SECI does not address.

**Habermas’s Theory of Communicative Action** introduces validity claims (truth, rightness, sincerity, comprehensibility) that Signal Theory does not formalize. Comprehensibility maps to Genre competence and Structure. Truth and sincerity are properties of the *content* of the Signal, not the architecture of the communication system—Signal Theory, like Shannon’s original theory, optimizes transmission fidelity, not truth (Section 11.4). Rightness (normative validity) is partially captured by Beer’s System 5 (Policy) but not fully formalized. This is a genuine gap: Signal Theory does not model the normative dimension of communication. Whether this constitutes a missing constraint or a deliberate scope limitation depends on whether normative validity failures produce communication system failures that the four constraints cannot diagnose—an open question.

Signal Theory’s contribution is architectural, not mathematical—it integrates existing constraints into a unified framework and applies them to the problem none of the individual theories addresses alone. The broader historical trajectory reveals a progression of fundamental questions—from Turing’s *can it compute?* (Turing, 1936) through Shannon’s *can it transmit?* (Shannon, 1948), Wiener’s *can it self-correct?* (Wiener, 1948), Ashby’s *can it regulate?* (Ashby, 1956), and Beer’s *can it survive?* (Beer, 1972). Each generation’s question subsumes the previous: a system that communicates optimally must also be viable, must regulate variety, must self-correct, must transmit reliably, and must compute.

### 11.3 Measurement

Signal Theory is being applied by the author in active organizational environments—mapping Signal flows against the taxonomy, diagnosing governing principle violations using the failure modes (Section 6.4), implementing corrective actions, and measuring the proxy metrics defined below before and after intervention. The methodology follows a diagnostic protocol: (1) catalogue the organization’s Signal types across all five dimensions, (2) compute the requisite variety ratio, (3) identify governing principle violations, (4) implement targeted corrective actions, and (5) measure proxy metrics to assess impact. Results from these implementations will be formally published; the measurement instruments are defined here to enable independent replication.

This paper applies Shannon’s principles *structurally*—the architecture and the existence of optimal encoding schemes are applied directly, while the quantitative equation provides directional guidance. S/N ratio is computable in digital channels (telecommunications, software systems, data pipelines) where bandwidth and noise are measurable quantities; in organizational channels (team communication, strategic alignment), S/N is a directional constraint—the principle that every channel has a finite ceiling—operationalized through the proxy metrics below rather than computed directly. The following proxy metrics bridge theory to measurement:

Metric	What It Measures	Maps To
Action Completion Rate	% of Signals that result in completed action items	Fidelity
Re-encoding Frequency	Count of follow-up messages re-explaining intent	Noise level
Time-to-Decode	Time from Signal receipt to first meaningful action	Decoding efficiency
Signal Bounce Rate	% of Signals re-routed after initial receipt	Routing accuracy
Genre Recognition Rate	Whether the receiver responds with appropriate genre convention	Genre competence
Feedback Loop Closure Rate	% of Signals with confirmed feedback response	Feedback completeness

**Table 17:** Proxy metrics bridging theory to measurement.

These metrics are *observable*, *measurable*, and *actionable*. Any organization already has these data points—every re-encoding event (“wait, what did you mean?”) is measurable today. Formal publication of results from active implementations is underway.

### 11.3.1 Quantifying Requisite Variety

Ashby’s Law can be operationalized as a measurable ratio:

$$R = \frac{V(\text{system})}{V(\text{disturbance})} \quad (3)$$

Where  $V(\text{system})$  is the variety of Signal types, genres, modes, and structures the system can produce, and  $V(\text{disturbance})$  is the variety of communication situations the system encounters. The ratio provides a diagnostic:

Ratio	Assessment	Implication
$R < 1.0$	<b>Failing</b> —system cannot handle the situations it encounters	Immediate variety expansion needed—new genres, modes, structures
$R 1.0\text{--}3.0$	<b>Marginal</b> —system handles most situations with strain	Targeted expansion for gap areas
$R 3.0\text{--}10.0$	<b>Adequate</b> —system handles situations with room for growth	Monitor for emerging variety gaps
$R \geq 10.0$	<b>Robust</b> —system handles encountered variety with significant surplus	Stable; prune unused variety to reduce complexity

**Table 18:** Requisite variety ratio diagnostic.

$V(\text{system})$  is enumerated by cataloguing the distinct genre-mode-framework combinations the organization actively employs.  $V(\text{disturbance})$  is enumerated by sampling communication situations over a defined period and classifying each by the genre-mode-framework combination it requires. The ratio is approximate—variety in Ashby’s original formulation counts distinguishable states, and the granularity of classification affects the count—but the diagnostic value lies in the relative magnitude, not the absolute precision.

An organization that has one genre for all communication (the “everything is a Slack

message” pathology) has  $R$  approaching 1.0 for a complex environment. An organization with established genres for every recurring communication situation—specs, briefs, reviews, reports, retrospectives, handoffs—has  $R$  proportional to its genre repertoire.

### 11.3.2 Triple-Layer Signal Verification

Signal quality can be assessed across three independent validation dimensions, each capturing a different aspect of encoding effectiveness:

**Formal verification (40% weight):** Does the Signal satisfy its structural requirements? Preconditions met, logical consistency maintained, goal state achievable from the encoded content. A spec that references undefined requirements fails formal verification. A roadmap with contradictory timelines fails formal verification.

**Semantic verification (35% weight):** Does the Signal preserve intended meaning? This is measured by correspondence between the encoder’s intent and the decoder’s interpretation—not string-level matching but meaning-level matching. “Approved” and “looks good to me” are semantically equivalent despite lexical difference. A Signal with high formal validity but low semantic correspondence has correct structure but wrong meaning.

**Information-theoretic verification (25% weight):** Does the Signal reduce uncertainty? Measured as entropy reduction—the difference between the receiver’s uncertainty before and after decoding. A Signal that tells the receiver nothing they did not already know has zero information-theoretic value regardless of formal and semantic quality. A Signal that resolves the key open question has high information-theoretic value even if its formal structure is imperfect.

The three layers are ordered by dependency: formal validity is a precondition for semantic validity (a structurally incoherent Signal cannot preserve meaning), and semantic validity is a precondition for information-theoretic value (a meaningless Signal cannot reduce uncertainty). This dependency ordering establishes which layer to check first—not the relative magnitude of each layer’s contribution to quality, which is an empirical question that varies by domain. A representative weighting for illustration:

$$\text{Quality} = w_1 \times \text{formal} + w_2 \times \text{semantic} + w_3 \times \text{info-theoretic} \quad (4)$$

where  $w_1 > w_2 > w_3$  and  $w_1 + w_2 + w_3 = 1$ .

The threshold for effective communication is empirically determinable per domain. The framework’s contribution is the decomposition into three independently measurable dimensions, not the specific weights.

## 11.4 Scope and Boundary Conditions

Signal Theory applies to any system where intent is encoded, transmitted through a channel, and decoded into action—organizational communication, software systems (APIs, message queues, event architectures), biological systems (nervous system, immune response), and educational systems (instruction, assessment, feedback). The framework is most powerful when communication is **purposive** (intended to produce action) and **structured** (amenable to genre and framework analysis). It is less applicable to purely

expressive, aesthetic, or ambient communication where the “intent → action” chain is not the primary function.

The framework identifies **encoding cost**—the time, attention, and resources required to produce an optimally encoded Signal—as a constraint the framework itself explains. The encoder is a node in the Signal Network, and encoding is itself a Signal processing operation bounded by the encoder’s own channel capacity. A manager with 30 minutes before a launch has lower encoding bandwidth than one with three hours; the optimal encoding given the encoder’s capacity constraint differs from the optimal encoding given unlimited time. When encoding bandwidth is insufficient, the framework predicts the consequence: the encoder defaults to lower-variety encoding (one brief for all receivers instead of four genre-specific briefs), producing a downstream Ashby violation that manifests as re-encoding, clarification requests, and misalignment—failures whose cumulative cost often exceeds the encoding investment that would have prevented them. The trade-off is not between quality and no cost; it is between encoding cost *upstream* and failure cost *downstream*. Dual-process encoding (Section 10.6) provides the primary mitigation: established genres convert expensive System 2 encoding into near-free System 1 encoding, reducing the marginal cost of high-quality encoding over time.

The recursion is bounded. At the level of a single utterance, the architectural principles (Beer, Wiener) become constraints on the *system that produces* the utterance rather than on the utterance itself. The framework’s recursion applies meaningfully from the level of a team’s communication patterns upward; below that threshold, only the encoding principles (Mode, Genre, Type, Format, Structure) and Shannon’s channel constraint apply.

The framework’s core model—encoder → Signal → decoder—is structurally monologic: it traces the path from one intent to one action. This is a deliberate analytical simplification, not an ontological claim. Bakhtin’s dialogism establishes that meaning is not transmitted intact but co-constructed between encoder and decoder—every utterance carries the traces of prior utterances and anticipates the response of the addressee. Hall’s encoding/decoding model (Hall, 1973) provides the complementary mechanism: the same Signal receives dominant, negotiated, or oppositional readings depending on the decoder’s position. Signal Theory accounts for interpretive variance through the feedback loop (Wiener) and the fidelity metric (the gap between encoded intent and decoded interpretation), but it does not model the interpretive process itself. The framework optimizes the conditions under which intended meaning is most likely to survive transmission; it does not claim to eliminate interpretive variance, which is irreducible.

Like Shannon’s original theory, Signal Theory optimizes transmission fidelity, not truth—a framework that only applies to true messages would not be a theory of communication. Decoder attention is modeled as bandwidth at the cognitive level; when bandwidth decreases, capacity decreases (Bandwidth Overload, Section 6.4). Power dynamics—political distortion, strategic ambiguity, gatekeeping—are modeled as Adversarial Noise in the failure taxonomy (Section 6.4). The genre conventions in this paper reflect primarily Western, English-language, technology-industry norms; the five-dimensional taxonomy is culturally extensible—the *dimensions* are universal even when the *specific genres* vary. A Japanese *ringi-sho* occupies the same dimensional space as a Western RFC.

## 11.5 Testable Predictions

The framework generates specific, falsifiable predictions from its governing principles. Each prediction derives from an established theorem or principle applied to organizational communication; the predictions are the empirical test of that application.

**P1 (Shannon):** When Signal volume exceeds a receiver’s processing bandwidth—measured by re-encoding frequency and time-to-decode—action completion rate will decrease. Reducing volume to below the receiver’s measured capacity will restore completion rates. *Falsified if:* volume reduction shows no effect on action completion.

**P2 (Ashby):** Organizations with higher requisite variety ratios ( $R > 3.0$ ) will show fewer genre mismatch failures and higher first-decode success rates than organizations with  $R < 1.5$ , controlling for organization size and industry. *Falsified if:* variety ratio shows no correlation with mismatch frequency.

**P3 (Beer):** Removing or lacking a specific VSM subsystem produces a specific, predictable failure mode: missing System 2 (Coordination) produces cross-team misalignment; missing System 4 (Intelligence) produces blindsided-by-change failures; missing System 5 (Policy) produces inconsistent encoding standards. The failure type is determined by the missing subsystem, not by random variation. *Falsified if:* subsystem absence produces no degradation, or degradation is random rather than subsystem-specific.

**P4 (Wiener):** Adding confirmation steps to Signals that currently lack feedback loops will reduce re-encoding frequency and increase action completion rate, controlling for Signal complexity. *Falsified if:* feedback loop closure shows no effect on re-encoding frequency.

**P5 (Taxonomy Completeness):** The five-dimensional taxonomy (Mode, Genre, Type, Format, Structure) is sufficient to classify any purposive communication—any communication intended to produce action at a destination. *Falsified if:* a dimension of purposive communication is identified that cannot be mapped to any of the five dimensions.

Three candidate sixth dimensions merit explicit examination: *temporality* (synchronous vs. asynchronous), *authority* (the power relationship between sender and receiver), and *audience size* (one-to-one vs. one-to-many vs. many-to-many). This paper’s position is that temporality is a channel property (captured in Format—a Live Session is synchronous, a Document is asynchronous), authority is a network property (captured in the Signal Network’s routing topology, Section VI), and audience size is likewise a network property (the number of destination endpoints). If this reduction is incorrect—if any of these dimensions produces classification failures that the five-tuple cannot capture—then P5 is falsified. This is a genuinely open empirical question.

**P6 (Encoding Investment):** When encoding quality improves (as measured by the six proxy metrics in Section 11.3), total system communication cost decreases—the upstream investment in encoding produces a larger downstream reduction in re-encoding, clarification, and rework. *Falsified if:* improved encoding quality shows no reduction in downstream communication cost. *Methodological caveat:* isolating encoding quality as the independent variable in organizational settings is difficult due to confounding factors (personnel changes, tool changes, workload variation). Controlled interventions (improving encoding for one team while holding conditions constant for a comparison team) would provide the cleanest test.

**An honest assessment of testability:** P3 and P4 are the most cleanly falsifiable—they predict specific, observable outcomes from specific interventions. P1 and P2 are falsifiable in principle but require measurement instruments (receiver processing bandwidth, requisite variety ratio) that are not yet validated. P5 is falsifiable in the Popperian sense (find a counterexample) but the reductive arguments above provide escape routes that an honest framework should acknowledge. P6 is weakened by confounding. The framework is more falsifiable than most framework papers—but less falsifiable than it would like to be. The predictions are genuine commitments, not hedged tautologies, but the measurement infrastructure required to test them rigorously does not yet exist. Building that infrastructure is the most important item in Future Work.

The framework’s validity rests on the conjunction of its governing principles—each derived from established theory (Shannon, Ashby, Beer, Wiener) but integrated and applied to organizational communication as a unified diagnostic for the first time. The predictions above provide the empirical test of that integration. A framework that cannot be falsified is a taxonomy; a framework that states what would falsify it is a theory.

## 11.6 Intent Encoding and Autonomous Agent Architecture

The preceding sections established that the Signal  $S = (M, G, T, F, W)$  encodes intent across five classifiable dimensions, that the Optimal System provides the seven-layer architecture for maintaining signal quality, and that the four governing principles constrain every viable communication system at every level of recursion. This section applies these formalizations to a specific class of problem: encoding organizational purpose into autonomous agents—systems that must carry intent as an encoded parameter rather than receiving it as real-time human guidance.

The problem is precise: when an autonomous agent operates without continuous human oversight, organizational intent must be encoded into the agent’s architecture as a Signal. If the encoding is inadequate—if the agent cannot decode organizational purpose from the signals it receives—the agent optimizes for what it can measure rather than what the organization needs. This is not a technology failure. It is a Signal encoding failure, and the four-principle diagnostic identifies exactly where the encoding breaks.

### 11.6.1 The Intent Encoding Problem as Governing Principle Violations

Deploying autonomous agents within an organization introduces three architectural requirements. Each maps to a governing principle, and failure to satisfy each produces a specific, diagnosable failure mode.

**Requirement 1: Unified Signal Infrastructure.** Every autonomous agent operates within a signal network—receiving inputs from organizational data sources, processing them, and transmitting outputs. When agents are deployed without a unified network topology (each team constructing its own context pipelines, its own data integrations, its own routing conventions), the result is fragmented channels with no coherent architecture. This is a **Shannon violation**: the organization’s total channel capacity is degraded by fragmentation, redundancy, and inconsistency. The Optimal System addresses this at Layer 1 (The Network) and Layer 5 (The Data Layer), which together define the architectural decisions—data governance, access controls, freshness guarantees, semantic consistency—that determine effective channel capacity. A protocol standard is neces-

sary but insufficient: the standard provides the physical layer, but the Optimal System provides the seven layers above it.

**Requirement 2: Governed Agent Repertoire.** When individual practitioners deploy agents independently—each with different capabilities, different workflows, and different interaction patterns—the organization’s variety of agent coordination patterns does not match the variety of tasks it faces. This is an **Ashby violation**: uncontrolled proliferation of agent behaviors creates variety in the coordination layer without corresponding variety in the governance layer. Signal Theory addresses this through the Agent node definition (companion paper, Section III), which establishes autonomous actors as first-class nodes in the signal network with defined properties: genre competence (what signal formats the agent can encode and decode), channel access (what protocols it supports), bandwidth (its compute capacity), and operator assignment (the human or institution liable for its behavior). The difference between individual agent use and organizational agent leverage is the difference between an isolated endpoint and a governed network.

**Requirement 3: Encoded Organizational Purpose.** The most critical requirement is that agents carry machine-readable expressions of organizational intent: goal structures, delegation frameworks, decision boundaries, escalation protocols, trade-off hierarchies, and feedback mechanisms. This requirement invokes all four governing principles simultaneously:

<b>Intent Encoding Requirement</b>	<b>Signal Theory Formalization</b>	<b>Principle</b>	<b>Location</b>
Machine-readable goals	<b>The Signal <math>S = (M, G, T, F, W)</math></b> —intent encoded into five classifiable dimensions, with Type (Searle’s illocutionary force) defining the action the signal demands	Shannon, Ashby	Section IV
Delegation frameworks	<b>Beer’s Viable System Model</b> —recursive five-subsystem architecture where each level handles only the variety appropriate to its scale, with System 5 (Policy) defining intent at every level of recursion	Beer	Section VIII, companion paper Layer 7

<b>Intent Encoding Requirement</b>	<b>Signal Theory Formalization</b>	<b>Principle</b>	<b>Location</b>
Decision boundaries	<b>Ashby's Law of Requisite Variety</b> —a mathematical theorem (not a heuristic) proving that the regulator's variety must equal or exceed the variety of disturbances it faces; decision boundaries are variety attenuation boundaries	Ashby	Section II
Escalation protocols	<b>Algedonic channels</b> —viability-preserving bypass signals that route directly from the point of disturbance to the policy level (System 5) when normal processing channels are too slow or too noisy	Beer, Shannon	Companion paper, Coordination Theory Section V
Trade-off hierarchies	<b>Variety engineering</b> —the dual mechanisms of attenuation (reducing environmental variety before it reaches the regulator) and amplification (increasing the regulator's variety to match what remains), managed at every system boundary	Ashby	Companion paper, Coordination Theory Section IV

<b>Intent Encoding Requirement</b>	<b>Signal Theory Formalization</b>	<b>Principle</b>	<b>Location</b>
Feedback mechanisms	<b>Wiener’s feedback architecture</b> —negative feedback loops at every level of recursion, with single-loop (did this output match intent?), double-loop (is our process producing good outputs?), and triple-loop (are our assumptions still valid?)	Wiener	Section II, companion paper Layer 6
Encoded judgment	<b>The Transposition of the Mind</b> —the theoretical foundation for encoding cognitive processes (reasoning patterns, decision frameworks, mental models) as signals that preserve their actionable meaning across the network	Shannon, Ashby	Section 2.1.1
Alignment drift detection	<b>Signal-to-Noise Ratio</b> —the root metric of the entire framework, measured through six proxy metrics (Section 11.3) and the triple-layer verification system (formal, semantic, information-theoretic)	Shannon	Section 11.3

**Table 19:** Intent encoding requirements mapped to Signal Theory.

The intent encoding problem is not a new problem requiring a new theory. It is the original problem of Signal Theory—encoding intent so that it survives transmission through noise and produces action at the destination—applied to a new class of destination node: the autonomous agent.

### 11.6.2 Diagnostic Application: The Autonomous Agent Failure Pattern

Consider a representative case: an organization deploys an autonomous customer service agent that handles millions of interactions per month across multiple markets and languages. Resolution times collapse from minutes to seconds. The organization projects significant cost savings. The agent performs the work equivalent of hundreds of human employees.

The agent is technically brilliant. It is also destroying the organization’s customer relationships. Customers report generic answers, robotic tone, and no ability to handle situations requiring judgment. The organization’s leadership eventually acknowledges that cost was the predominant evaluation factor and quality suffered. The organization begins rehiring the human agents it eliminated.

The naive reading is that AI cannot handle nuance. Signal Theory provides a precise diagnostic:

```
AUTONOMOUS AGENT FAILURE -- Signal Theory Four-Principle Diagnostic

Principle 1 -- Shannon (Channel Capacity):          SATISFIED
  The agent processed millions of conversations per month.
  The channel was adequate. No bandwidth violation.

Principle 2 -- Ashby (Requisite Variety):          VIOLATED
  The agent's response repertoire matched only one mode: efficient resolution.
  Customer situations requiring judgment, empathy, policy flexibility,
  retention gestures, or escalation had no corresponding signal type in
  the agent's genre competence. The agent's variety (V_agent) was far
  below the variety of customer emotional and situational states (V_customer).
  Result: V_agent < V_customer -> unregulable states -> customer dissatisfaction.

Principle 3 -- Beer (Viable Structure):            VIOLATED
  The agent operated as System 1 (Operations) only -- executing ticket
  resolution as its sole function. No System 4 (Intelligence) scanned
  customer sentiment or detected relationship degradation. No System 5
  (Policy) encoded the organizational intent that customer lifetime value
  outweighs resolution speed. The agent had no recursive viability --
  it was a single-function processor without the governance architecture
  required for autonomous operation.

Principle 4 -- Wiener (Feedback Closure):          VIOLATED
  No return loop connected customer satisfaction signals to agent behavior
  adjustment. The agent could not sense that its optimization for speed
  was destroying trust. The feedback path from customer experience to
  agent parameters was open -- errors accumulated uncorrected.

ROOT CAUSE: 3 of 4 governing principles violated simultaneously.
The agent optimized within its channel capacity (Shannon satisfied)
for the only objective it could measure (resolution speed) because
no one had encoded the objectives that mattered most: relationship
quality, brand trust, customer lifetime value, and the contextual
judgment about when to be efficient versus when to be generous.

SIGNAL THEORY CLASSIFICATION:
- Ashby violation -> Variety Deficit failure mode
- Beer violation -> Missing Subsystem failure mode (Systems 4 and 5 absent)
- Wiener violation -> Open Loop failure mode

CORRECTIVE ACTIONS:
```

1. Expand agent genre competence to include retention, escalation, and empathy signal types (Ashby remediation)
2. Implement System 4 (customer sentiment analysis) and System 5 (organizational value hierarchy encoded as decision boundaries) (Beer remediation)
3. Close the feedback loop: customer satisfaction metrics must feed back into agent behavior parameters in real time (Wiener remediation)

The human agents who were replaced carried the institutional knowledge that constituted Systems 4 and 5—the ability to sense customer emotional state (System 4 intelligence) and the internalized understanding of when to bend policy, when to spend extra time, when efficiency should yield to generosity (System 5 encoded judgment). This knowledge had never been formalized. It existed as tacit knowledge—what experienced practitioners “just know.” Signal Theory’s contribution is the architecture for making tacit knowledge explicit, structured, and machine-actionable—not as prose in a configuration file, but as governing constraints with mathematical foundations.

This failure pattern generalizes. Deploying an AI tool across an organization without encoding organizational intent is equivalent to hiring thousands of new employees and never telling them what the organization does, what it values, or how to make decisions. The tool may have adequate channel capacity (Shannon) but no variety to match the diversity of organizational tasks (Ashby), no recursive governance structure connecting tool behavior to organizational purpose (Beer), and no feedback loop from organizational outcomes to tool configuration (Wiener). The diagnostic is the same regardless of vendor, model capability, or deployment scale.

### 11.6.3 Agent Autonomy as Variety Transfer

Autonomous agents operate at varying levels of independence from human oversight. This variation is not merely a feature configuration—it is a **variety transfer** from operator to agent. Each increase in agent autonomy transfers governance variety from the human to the agent, and the agent’s internal architecture must possess the corresponding VSM subsystems to absorb that variety. Deploying an agent at high autonomy with low internal architecture is a guaranteed Ashby violation.

Signal Theory formalizes five levels of agent autonomy, each defined by the VSM subsystem scope the agent possesses:

Level	Operator Role	Agent’s VSM Scope	Governing Principle Requirements
L1: Operator	Human provides all direction. Agent executes specific tasks on demand.	System 1 (Operations) only.	Shannon: channel between human and agent must be adequate for instruction throughput. Minimal architectural requirements.

<b>Level</b>	<b>Operator Role</b>	<b>Agent's VSM Scope</b>	<b>Governing Principle Requirements</b>
L2: Collaborator	Human and agent share planning and execution with fluid control handoffs.	Agent has partial Systems 1–2 (Operations + Coordination). Human retains Systems 3–5.	Ashby: agent's coordination variety must match task variety within its scope. System 2 prevents oscillation between human and agent control.
L3: Consultant	Agent takes initiative. Human provides expertise, preferences, and directional guidance.	Agent has Systems 1–4 (Operations through Intelligence). Human retains System 5 (Policy).	Beer: agent requires autonomous operations with policy oversight. Algedonic channel (viability bypass) enables the agent to escalate viability threats directly to the human policy level.
L4: Approver	Agent operates independently. Human gates consequential actions.	Agent has Systems 1–5 with human veto at System 5 for high-stakes decisions.	Ashby: variety engineering at the System 5 boundary—attenuating the agent's decision variety to exclude decisions requiring human approval. The boundary must be mathematically defined, not informally specified.
L5: Autonomous	Agent has full operational independence. Human monitors via logs with emergency intervention only.	Agent has complete Systems 1–5. Human has emergency stop (algedonic channel from human to agent's System 5).	All four principles simultaneously. Without encoded organizational intent at System 5, the agent optimizes for measurable objectives that may destroy unmeasured objectives. This is the failure mode described in §11.6.2.

Level	Operator Role	Agent’s VSM Scope	Governing Principle Requirements
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**Table 20:** Five levels of agent autonomy as VSM subsystem scope.

The critical insight is that **autonomy level is not a permission setting—it is an architectural requirement**. An agent deployed at Level 5 without Systems 2–5 internally encoded is not “autonomous”—it is unregulated. The organization has transferred governance variety to an entity that cannot absorb it. Ashby’s Law is violated by construction, and the resulting failure modes (variety deficit, missing subsystem, open loop) are predictable from the framework.

**Autonomy certification** follows naturally: before an agent can be deployed at a given autonomy level, a **variety audit** must verify that the agent’s internal architecture possesses the VSM subsystems required for that level. An agent whose genre competence spans only routine resolution cannot be certified for Level 3, where judgment-requiring situations fall within its scope. An agent without System 4 (environmental scanning) cannot be certified for Level 4, where it must independently detect when to escalate. The certification is not a bureaucratic process—it is a mathematical verification: does the agent’s internal variety match the environmental variety it will face at this autonomy level?

#### 11.6.4 The Context Architecture Completeness Requirement

Current approaches to agent context management typically separate context into distinct layers: immediate working state, session-scoped memory, long-term organizational knowledge, and versioned artifacts. This four-layer decomposition addresses a subset of the Optimal System’s seven-layer architecture:

Typical Context Layer	OS Layer	What Is Missing
Working Context (immediate state, instructions, tool outputs)	Layer 2 (The Signal) + Layer 3 (Composition)	Typical approaches treat context as undifferentiated data. The Optimal System classifies every piece of context as a Signal $S = (M, G, T, F, W)$ with genre competence matching and illocutionary force.
Session Memory (interaction log within a conversation)	Layer 5 (The Data Layer)—session scope	Typical approaches capture events chronologically. The Optimal System structures data by Signal taxonomy with validation at capture and explicit DIKW transformation traceability.

Typical Context Layer	OS Layer	What Is Missing
Long-Term Memory (persistent knowledge across sessions)	Layer 5 (The Data Layer)—organizational scope	Typical approaches provide searchable storage. The Optimal System provides the full DIKW hierarchy (Data → Information → Knowledge → Wisdom) with each transformation explicit.
Artifacts (versioned binary/textual data)	Layer 5 (The Data Layer)—artifact management	Typical approaches address artifacts by name and version. The Optimal System governs artifacts through the full Signal architecture including routing rules, genre competence at destinations, and feedback mechanisms.
<i>Typically absent</i>	Layer 1 (The Network)	No network topology—no node types, no routing rules, no relay distortion prevention, no cross-subnet redundancy.
<i>Typically absent</i>	Layer 4 (The Interface)	No interface theory—no progressive disclosure, no receiver bandwidth matching, no noise reduction.
<i>Typically absent</i>	Layer 6 (The Feedback Loop)	No feedback architecture—no single/double/triple-loop framework, no fidelity measurement, no alignment drift detection.
<i>Typically absent</i>	Layer 7 (The Governance)	No governance architecture—no VSM, no recursive viability, no System 5 policy encoding, no algedonic bypass for viability threats. This is the most critical absence.

**Table 21:** Context architecture completeness: typical four-layer vs. Optimal System seven-layer.

The four-layer approach addresses the data management substrate of agent operation. It does not address the communication architecture (Layer 1), the interface through which agents present information to humans (Layer 4), the feedback mechanisms that detect and correct misalignment (Layer 6), or the governance architecture that encodes organizational purpose at every level of recursion (Layer 7). It is Layer 5 without Layers 1, 4, 6, and 7—and it treats Layers 2 and 3 as undifferentiated “context” rather than

as five-dimensionally classified Signals with genre competence, illocutionary force, and compositional micro-structure.

The completeness requirement is not optional. An agent context architecture missing Layer 6 (Feedback) cannot detect alignment drift. An architecture missing Layer 7 (Governance) cannot encode organizational purpose. An architecture missing Layer 1 (Network) cannot reason about signal routing, relay distortion, or redundancy. The seven-layer Optimal System is the minimum complete architecture for autonomous agent operation—any subset produces the failure modes catalogued in Section VI §6.4 and the companion paper Section X.

### 11.6.5 Why Intent Encoding Is Difficult

Three structural factors explain why encoding organizational intent into autonomous agents remains an unsolved problem in practice:

**First, the problem is new in its applied form.** Before agents could run autonomously over long time horizons, the human was the intent layer. The human stood between the model and the organization’s purpose, translating organizational intent into instructions, evaluating outputs against unstated criteria, and correcting misalignment in real time. Long-running autonomous agents break this model: the human is no longer present at every decision point, and the agent must carry organizational intent as an encoded parameter rather than receiving it as real-time guidance. The Transposition of the Mind (Section 2.1.1) formalizes this requirement: cognitive processes—reasoning patterns, decision frameworks, mental models—can be encoded as signals that preserve their actionable meaning across the network. The question is not whether such encoding is possible but whether organizations have done the work of making their intent explicit.

**Second, organizational strategy and agent engineering are separated by a disciplinary gap.** The people who understand organizational purpose (executives, strategists) are not the people who build agents (engineers, ML researchers), and the people who build agents do not typically understand organizational strategy. This separation guarantees an intent encoding gap: engineers can build infrastructure, but intent comes from the entire leadership team working together. Beer’s VSM provides a shared language—the five subsystems, the recursive architecture, the algedonic channel—that both organizational thinkers and technologists can use to reason about the same system.

**Third, making organizational intent explicit is extraordinarily difficult.** Most organizations have never had to do this because humans filled the gap. Goals live in slide decks, OKR documents that get half-read, leadership principles that get cited in performance reviews but never operationalized, and the tacit knowledge of experienced employees who know what to do in ambiguous situations even though they have never been told. Signal Theory’s five-dimensional taxonomy provides the encoding framework: organizational intent is not a single parameter but a Signal  $S = (M, G, T, F, W)$  where the Mode defines how intent is represented, the Genre defines the conventions for that representation, the Type defines what action the signal demands, the Format defines the container, and the Structure defines the internal skeleton. Converting organizational objectives into agent-actionable parameters is a Signal encoding problem—and Signal Theory provides both the encoding architecture and the failure modes that diagnose when encoding is inadequate.

The trajectory from individual instruction through information environment to organizational purpose encoding is a convergent discovery of the problem space that Signal Theory formalizes as a unified architecture. The Optimal System is the target architecture. The four governing principles are the constraints. The failure taxonomy is the diagnostic. The seven layers are the implementation specification. Intent encoding is Signal Theory applied to a new class of destination node.

## 12 Future Work

Signal Theory is recursive in its own application: the work that remains is itself a Signal Theory problem. The interfaces, tools, measurement systems, and expanded taxonomies described below do not yet exist—because the Signals that would create them have not yet been encoded. Each item below is a Signal that must be composed across Mode, Genre, Type, Format, and Structure before it can be transmitted, decoded, and acted upon. The future of Signal Theory is, in the most literal sense, a set of Signals waiting to be created.

1. **Formal publication of empirical results**—Signal Theory is being applied by the author in organizational environments. The proxy metrics defined in Section 11.3 are under initial measurement. The next step is formal publication—encoding the findings into academic genre conventions (peer-reviewed paper, case study)—itself a Signal encoding problem.
2. **The Optimal System Interface**—The framework describes the architecture; the interface through which users would experience it does not yet exist. The next interface must be *native to Signal Theory*—presenting the five dimensions, diagnosing failures in real time, and assisting encoding. The interface is downstream of the Signal; the Signal must come first.
3. **Computational classification engine**—The five-dimensional taxonomy is structured enough to be machine-readable. A Signal’s dimensions could be automatically classified based on intent and receiver competencies, detecting genre mismatches and flagging failure modes before transmission—a real-time application of the failure taxonomy (Section 6.4).
4. **Cross-cultural genre catalogue**—Cataloguing genre conventions across cultures—Japanese *ringi-sho*, German *Protokoll*, Brazilian *ata de reunião*—would test the taxonomy’s universality (Section 11.4). Each culture’s genre repertoire either confirms or challenges the dimensional model.
5. **Genre lifecycle dynamics**—Genres emerge (the tweet thread, the Loom walk-through), evolve (memo → email → Slack message), and die (the telegram, the fax cover sheet). Modeling this lifecycle would extend the framework’s temporal dimension and provide predictive power for genre emergence.
6. **Recursive organizational implementation**—Mapping an organization’s actual Signal flows against Beer’s VSM, identifying where communication fails to satisfy the governing constraints, and restructuring accordingly. Theory becomes practice—not as a test of whether it works, but as the application of principles already established in Section II.

7. **Augmented intelligence as human-machine signal architecture**—Signal Theory formalizes how communication works. The next step is formalizing what happens when a human encoder is augmented by a machine system that handles recall—data retrieval, scenario generation, research assembly, pattern matching—while the human handles deep thought—reasoning, judgment, synthesis, strategic decision-making. This is not artificial intelligence (replacing the human encoder with a machine encoder). It is augmented intelligence (extending the human encoder’s channel capacity by offloading recall to a system whose retrieval bandwidth approaches channel capacity for structured data). The split is precise: human cognition excels at instant pattern recognition from compressed experience—what the companion cognition papers formalize as “hot constructs.” Machine systems excel at what Shannon’s source coding theorem demands—lossless retrieval of structured information across unbounded repositories. Together, they constitute a signal-processing architecture where the human and machine each operate in their zone of maximum S/N. The Optimal System’s Layer 7 (Governance) provides the viability architecture. The augmented intelligence architecture is the applied endpoint of Signal Theory: every principle in this paper governs how the augmentation must be structured to maximize Signal-to-Noise Ratio at the human-machine boundary.

## 13 Conclusion

Signal Theory unifies information theory, cybernetics, speech act theory, and multimodal semiotics into a single architecture for optimal communication. It makes five contributions:

1. The **Signal**—the fundamental unit of actionable communication, classified across five dimensions (Mode, Genre, Type, Format, Structure), resolving the naming fragmentation across disciplines.
2. The **Optimal System**—a cybernetic architecture derived from four governing principles (Shannon, Ashby, Beer, Wiener) that any viable communication system must satisfy at every level of recursion.
3. The **Path of Least Resistance**—not a metaphor but a consequence of Shannon’s channel coding theorem, Zipf’s Principle of Least Effort, and the empirical evidence of Morse code. Encoding quality is a probabilistic investment whose returns compound over time.
4. **Technology-independence**—Shannon’s architecture has remained constant across every major interface shift from Morse code to AI-assisted tools. Signal Theory classifies at the level of communicative function, not technological implementation.
5. **Machine Intelligence as implementation layer**—The Optimal System described by this framework approaches full achievability for the first time as Machine Intelligence bridges siloed interfaces, translates between genre conventions, and computationally optimizes encoding across the entire Signal network—from strategic directives down to the most granular unit of daily work.

The framework provides eleven distinct failure modes with corrective actions, six proxy metrics for measuring Signal quality, a requisite variety ratio for diagnosing system-level

adequacy, and a design methodology that applies recursively from a single message to an entire organization.

The four governing principles—Shannon, Ashby, Beer, Wiener—constitute the cybernetic foundation of Signal Theory. They describe how signals are encoded, transmitted, decoded, and fed back through channels of bounded capacity. But signals do not propagate through static, linear, equilibrium environments. They propagate through environments that are **chaotic** (Lorenz, 1963)—where small perturbations in encoding produce disproportionate effects on reception, and where the same signal transmitted at slightly different times can produce entirely different outcomes. They propagate through environments that are **reflexive** (Soros, 1987; Giddens, 1984)—where measuring a signal changes the signal, where observing behavior alters behavior, and where the act of communication restructures the communicative environment itself. And they propagate through environments that are **complex adaptive** (Holland, 1995)—where receivers learn, encoders evolve, channels mutate, and the signal landscape co-evolves with the agents who inhabit it. These three extensions—chaos dynamics, reflexive theory, and complex adaptive systems theory—do not replace the cybernetic foundation. They reveal its boundary conditions. Shannon’s channel capacity holds in chaotic regimes, but the effective S/N ratio fluctuates nonlinearly. Ashby’s requisite variety holds in adaptive systems, but the disturbance variety evolves as agents co-adapt. Beer’s recursive viability holds in reflexive environments, but the system’s own monitoring signals perturb the system being monitored. Wiener’s feedback holds everywhere, but in reflexive systems the feedback *constitutes* part of the signal it measures. Three companion papers formalize these extensions: *Chaos Dynamics and Viability Theory* (nonlinear dynamics in signal systems), *Adaptive Systems Theory* (complex adaptive systems and emergence), and *Reflexive Economics* (observer effects in economic signal systems). Signal Theory is the root. Chaos, adaptation, and reflexivity are the soil conditions in which it grows.

The Signal was always there—traveling through every channel, encoded in every message, decoded at every destination, persisting through chaos, adapting through reflexivity, emerging through complexity. This paper gives it a name, a taxonomy, and an architecture. The companion papers give it the soil conditions in which it grows.

## A Worked Examples

Each Signal requires simultaneous choices across all five dimensions:

```
SIGNAL: "Content Creator Brief for Q2 Campaign"  
|  
+-- MODE: Linguistic + Visual (written doc with reference images)  
+-- GENRE: Brief  
+-- TYPE: Direct (compels content creation)  
+-- FORMAT: Google Doc  
+-- STRUCTURE: Objective -> Audience -> Key Messages -> Tone ->  
                Deliverables -> Timeline
```

```
SIGNAL: "Architecture Decision for Auth System"  
|  
+-- MODE: Linguistic (written)  
+-- GENRE: ADR (Architecture Decision Record)
```

```

+-- TYPE: Inform (states the decision) + Direct (compels implementation)
+-- FORMAT: Markdown file in repository
+-- STRUCTURE: Context -> Decision -> Consequences -> Alternatives
                Considered

```

**Non-organizational examples**—the same architecture applies universally:

```

SIGNAL: "Pain Reflex Arc" (Biological)
|
+-- MODE: Electrochemical (nerve impulse)
+-- GENRE: Alarm (danger signal with standardized response)
+-- TYPE: Direct (compels withdrawal action)
+-- FORMAT: Action potential sequence
+-- STRUCTURE: Stimulus -> Sensory neuron -> Interneuron ->
                Motor neuron -> Response

```

```

SIGNAL: "REST API Response" (Software)
|
+-- MODE: Code (structured data)
+-- GENRE: Status Report (machine-readable state description)
+-- TYPE: Inform (states current system state)
+-- FORMAT: JSON over HTTP
+-- STRUCTURE: Status Code -> Headers -> Body -> Error Object ->
                Pagination

```

## B Key Definitions

Term	Definition
Signal	An encoded unit of intent designed to produce action at the destination—the payload that travels through noise
Mode	The sensory channel through which a Signal is perceived (linguistic, visual, auditory, gestural, spatial)
Genre	The conventionalized, socially recognized form of communication (SOP, proposal, brief, contract, spec, ADR)
Type	The communicative function a Signal performs (Direct, Inform, Commit, Decide, Express—mapping to Searle’s directive, assertive, commissive, declarative, expressive)
Format	The physical or digital container/vessel that carries the Signal (PDF, video, slide deck, email, repository)
Structure	The internal skeleton/template that organizes content within a genre
Package	A composed bundle of multiple Signals delivered together
Noise	Any interference between encoding and decoding that degrades the Signal

Term	Definition
Channel	The transmission medium through which a Signal travels
Encoding	The process of structuring intent into a transmittable form across all five dimensions
Decoding	The process of extracting actionable meaning from a received Signal
Channel Capacity	The maximum rate of reliable communication through a given channel ( $C = B \times \log_2(1 + S/N)$ )
Signal-to-Noise Ratio (SNR)	The ratio of useful information to interference in a communication
Entropy	The information content / surprise value of a message
Redundancy	Additional structure added to a Signal to ensure accurate decoding despite noise
Fidelity	The degree to which the decoded message matches the encoded intent
Genre Competence	A receiver's ability to decode a particular genre based on familiarity with its conventions
Optimal System (OS)	A communication architecture operating at maximum Signal-to-Noise Ratio across all dimensions
Node	A source, destination, or relay point in the Signal network—the organizational unit where encoding or decoding occurs
Network	The topology of connected Nodes through which Signals are routed
Signal Composition	The mode-specific micro-structure of a Signal (e.g., words → sentences → paragraphs for linguistic mode)
Signal Chain	A sequence of dependent Signals directed toward a bounded outcome (a project)
Signal Lifecycle	Created → Sent → Received → Decoded → Acted Upon
Data	Raw, unprocessed facts without context—the symbols before encoding
Information	Data + context + meaning—the decoded Signal
Knowledge	Accumulated decoded Signals—patterns recognized across many signals
Wisdom	The ability to encode optimal Signals based on accumulated knowledge and judgment
Interface	The display/interaction layer that decodes and presents information to human endpoints
Machine Intelligence	Non-human Signal processing nodes—AI agents, algorithms, ML models, data pipelines, real-time processors—that encode, decode, translate, and route Signals, reducing cognitive load on human endpoints
Feedback Loop	Circular causal path where the destination's response returns to the source for error correction
Viable System	A system organized to survive in a changing environment—requires 5 recursive subsystems (Beer)

Term	Definition
Requisite Variety	The principle that a controller must have at least as much complexity as the system it manages (Ashby)
Requisite Variety Ratio	$V(\text{system})/V(\text{disturbance})$ —the quantified measure of whether a system’s Signal repertoire matches its environmental complexity. $R < 1.0$ = failing; $R \geq 10.0$ = robust
Recursive Structure	Each subsystem of a viable system is itself a viable system with the same architecture
Autopoiesis	A system that continuously regenerates itself from within—monitors its own Signal complexity, detects degradation trajectories, and restructures before failure (Maturana & Varela, 1972; applied by Beer)
Syntegrity	Beer’s non-hierarchical coordination topology—icosahedral structure enabling parallel multi-topic decision-making with zero structural hierarchy
Herniation	A cross-layer integrity violation where a Signal bypasses architectural boundaries—a function skipping policy, a data request ignoring authority chains—producing structurally invalid communication
Dual-Process Encoding	System 2 (deliberative) encoding for novel Signals vs. System 1 (reflexive) encoding for recognized patterns. The Optimal System learns to convert System 2 to System 1 through genre establishment (after Kahneman, 2011)
Signal Topology	The temporal structure of Signal flow—sequential (each Signal depends on the previous) or parallel (independent Signals transmitted concurrently)
Triple-Layer Verification	Composite quality assessment: formal (structural validity, 40%) + semantic (meaning preservation, 35%) + information-theoretic (entropy reduction, 25%)

**Table 22:** Key definitions.

## C The Evolution of the Interface

Shannon’s architecture—Source → Encoder → Channel → Decoder → Destination—has never changed. What has changed is the sophistication of the decoding surface: written text (~3200 BCE) enabled persistent storage; the printing press (1440s) enabled one-to-many transmission; the telegraph (1844) made the channel electrical; telephone, radio, and television added synchronous auditory and broadcast modes; the GUI (1984) added visual and spatial logic; the web (1993) networked Signals; smartphones (2007+) made channels ubiquitous and multimodal; modern SaaS tools specialized interfaces by genre; AI-assisted interfaces (2020s+) augment encoding and decoding computationally; agent-connected systems (emerging) dissolve the boundaries between siloed interfaces.

Every major communication technology changed the Interface layer while leaving Shannon’s architecture intact. Signal Theory classifies at the level of communicative function precisely because this level persists across every technological shift.

## D Diagnostic Walkthrough

The following walkthrough illustrates how the four governing principles function as a diagnostic framework when applied to a common organizational failure. The scenario is representative; implementations using this methodology are currently underway.

### Scenario: Product Launch Communication Failure

A SaaS company prepares a product launch. The product team writes a detailed launch brief covering features, positioning, and timeline. The brief is distributed to Marketing, Engineering, Sales, and Support via a shared document. On launch day, customer support cannot answer basic questions about the new feature. The sales team is positioning the product incorrectly to prospects. Marketing’s campaign messaging contradicts the product team’s intended positioning. Engineering deployed correctly—the only team that executed as intended.

#### Step 1: Apply the Four Principles

**Shannon—Is the channel adequate?** The brief was distributed through a shared document accessible to all teams. Channel capacity was not exceeded. This is not a bandwidth problem.

**Bakhtin—Is the genre recognized?** The product team sent a single *launch brief* to all four teams. Marketing needed a *campaign brief* with positioning angles and audience segments. Sales needed an *enablement brief* with competitive differentiators and objection handling. Support needed a *runbook* with troubleshooting steps and escalation paths. Engineering needed a *spec* with implementation requirements. One genre was sent where four were required. **Variety failure**—the system’s genre repertoire did not match the variety of the receivers’ decoding requirements (Ashby violation).

**Beer—Is the architecture viable?** The launch coordination operated at System 3 (Control—the product team managing the launch) without System 2 (Coordination—a mechanism ensuring that all teams decoded the launch intent consistently). Each team decoded independently with no cross-team alignment. **Architecture failure**—missing System 2.

**Wiener—Is the feedback loop closed?** The brief was broadcast. No confirmation step verified that each team decoded the launch intent correctly. No mechanism existed for teams to signal back “here is what we understood” before launch day. The encoding error compounded through three weeks of independent preparation. **Feedback failure**—open loop.

## Step 2: Classify the Failure

This is a **triple failure**: Variety (Section 6.4—one genre for four distinct receivers), Architecture (Beer—missing System 2 coordination), and Feedback (Wiener—no confirmation loop). No single theory can produce this diagnosis. Shannon analysis alone would conclude “the channel is fine.” Bakhtin analysis alone would identify the genre mismatch but not the architectural or feedback failure. Only the integrated framework classifies all three failure modes simultaneously and prescribes distinct corrective actions for each.

## Step 3: Corrective Action

Failure	Principle Violated	Corrective Action
Variety	Ashby	Encode separate Signals for each team: Support Runbook, Sales Enablement Brief, Marketing Campaign Brief, Engineering Spec—each in the genre the receiver has competence to decode
Architecture	Beer	Add a System 2 coordination mechanism: a cross-functional sync (weekly for 3 weeks pre-launch) where each team confirms their decoded interpretation against the product team’s intent
Feedback	Wiener	Add a confirmation step: each team lead produces a one-page “decoded intent” summary that the product team reviews before launch preparation begins

**Table 23:** Corrective actions for the product launch failure.

## Step 4: Measurement

The proxy metrics from Section 11.3 provide the measurement framework:

- **Re-encoding Frequency:** Count of follow-up messages re-explaining launch intent after the brief is distributed. A decrease after corrective action indicates improved encoding fidelity.
- **Action Completion Rate:** Percentage of launch tasks completed as intended by each team. The baseline (failed launch) is the comparison point.
- **Genre Recognition Rate:** Whether each team responds within the conventions of their domain-specific genre—Support producing runbook-structured responses, Sales producing pitch-structured responses.
- **Feedback Loop Closure Rate:** Percentage of teams that confirmed their decoded interpretation before launch preparation.

The framework does not guarantee successful launches—it identifies the *specific failure modes* that caused the previous failure and provides *targeted corrective actions* whose effectiveness is measurable through the proxy metrics defined above.

*“The fundamental problem of communication is that of reproducing at one point either exactly or approximately a message selected at another point.”*

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