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ITEM #99 — Comprehensive Calling Graph Language (CCGL) and
Algorithmic Coverage Completeness

Conversation Title: 自主进化设计评析

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ITEM #99 — Comprehensive Calling Graph Language (CCGL) and Algorithmic Coverage Completeness

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1. Overview

The **Comprehensive Calling Graph Language (CCGL)** is proposed as the unified representational language for the **TaskLayer–ActionLayer** structure in the Digital Brain Model (DBM).

It integrates **Comprehensive Calling Graph (CCG)** and **Dynamic Comprehensive Calling Graph (DCGG)** as the structural framework through which **AI achieves autonomous algorithmic evolution**.

Its design principle asserts that:

Any algorithm yet to be written is a compositional combination of existing language elements.

This statement establishes the **Algorithmic Coverage Completeness (ACC)** principle — the structural counterpart of Turing completeness within semantic combinatorics.

2. The BUS Abstraction

To unify module interaction and abstract away implementation frictions, all operations communicate via a standardized interface:

```
public int operationXX(HashMap<String, Object> runtimeDataBus);
```

This **semantic BUS** serves as a universal I/O medium between programs and algorithmic modules.

Through this, every computational component becomes composable under a shared signature, allowing higher-level reasoning systems (like CCG/DCGG) to focus solely on the **structural composition** of operations.

Key Advantages:

- Interface unification and composability
- Contextual traceability of runtime data
- Differential analysis via BUS deltas
- Structural optimization via CCG traversal

3. From Operations to Algorithmic Language

Every program or statement is modeled as:

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3. From Operations to Algorithmic Language

Every program or statement is modeled as:

OpX -> [Op1, Op2, ..., OpN]

Examples:

- A single statement: $OpX \rightarrow []$
- A sequential algorithm: $OpX \rightarrow [Op1, Op2, Op3]$
- A nested composition: $OpX \rightarrow [Op1, [Op2a, Op2b], Op3]$

Thus, **CCGL** defines a **hierarchical combinatorial syntax**, similar in spirit to human language, where:

Sentence $\rightarrow [Word0, Word1, Word2, \dots]$

WordSequence $\rightarrow [Word / WordSequence[0], Word / WordSequence[1], \dots]$

This analogy reveals a deep structural isomorphism between **linguistic expressions** and **algorithmic expressions**.

4. Hierarchical Layers of CCGL

Level	Structure	Description
L0 — Atomic Operation	$OpX \rightarrow []$	Fundamental computation unit
L1 — Sequential Structure	$OpX \rightarrow [Op1, Op2, Op3]$	Ordered execution
L2 — Nested Composition	$OpX \rightarrow [Op1, [Op2a, Op2b], Op3]$	Conditional or iterative logic
L3 — Conceptual Compression (CCC)	$[Op2, Op3] \rightarrow Op23$	Abstraction and pattern emergence

Together, these levels support both **syntactic representation** and **semantic generalization**, enabling algorithmic thought to be expressed and evolved within the same linguistic substrate.

5. CCG vs DCGG

Graph	Definition	Function
CCG (Comprehensive Calling Graph)	Static structural representation of calling relations	Captures existing algorithmic architecture

Graph	Definition	Function
DCGG (Dynamic Comprehensive Calling Graph)	Temporal and adaptive evolution of CCG	Models algorithmic self-extension and contextual recomposition

6. Enrichment Mechanisms

CCGL evolves through two complementary enrichment processes:

6.1. CCC Extraction Enrichment

Extract recurrent operation subsequences and assign them symbolic names:

$[Op2, Op3] \rightarrow Op23$

This mechanism compresses redundancy, forming a library of **Common Concept Cores (CCCs)** — the linguistic atoms of higher reasoning.

6.2. Generative Enrichment

Extend sequences by adding new operations on either side:

$[OpA, OpB, OpC] \rightarrow [OpX, OpA, OpB, OpC, OpY]$

This operation mimics **creative generation** — forming new algorithms by contextual expansion.

Together, these processes constitute the **Compression–Expansion Cycle** of algorithmic language evolution, analogous to the **learning–creativity duality** in human cognition.

7. Algorithmic Coverage Completeness (ACC)

Let S be the set of all CCGL statements.

Define a generative operator $G(S)$ that applies enrichment transformations to S .

If for every computable algorithm P , there exists a sequence $s \in G(S)$ structurally equivalent to P ,
then CCGL satisfies **Constructive Coverage Completeness (C³T)**:

C³T Theorem:

A language closed under CCC extraction and generative enrichment is algorithmically complete.

This theorem provides a **constructive proof of algorithmic universality** through compositional semantics, without recourse to symbolic rewriting or machine models.

8. Integration with DBM Layers

DBM Layer	Function	CCGL Mapping
Prompt Layer	Natural language input	CCC-level concepts
Task Layer	Algorithmic composition	L1–L2 structured statements
Action Layer	Execution and feedback	L0 atomic operations

This mapping allows DBM to **compile thought into action**:
from natural language → algorithmic graph → runtime behavior.

Thus, **autonomous programming** becomes a direct consequence of language evolution within CCG/DCGG space.

9. Implementation Roadmap

1. **CCGL Parser/Compiler** — transforms OpX syntax into CCG JSON or Java classes.
 2. **CCC Extraction Engine** — identifies frequent operation patterns to form reusable cores.
 3. **Generative Engine** — explores enriched combinations to synthesize new algorithms.
 4. **Differential Evaluator** — evaluates algorithmic fitness via DBM’s differential trees.
 5. **Evolution Loop** — continuously refines CCC and generative pathways toward self-improving architectures.
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10. Philosophical Reflection

Language is the mother of algorithms; algorithms are the embodiment of language.
When a language can autonomously evolve its algorithms,
thought itself attains structural self-generation.

The CCGL framework thus marks a transition from **symbolic programming** to **evolutionary linguistic computation**,
laying the foundation for **AI systems that write and refine their own operational language.**

