

University of Stuttgart Institute of Industrial Automation and Software Engineering

Chatbot

You:

In the balance population simulation process,what factors are related to the birth term and the death term?

AI:

Introduction

In the context of balance population simulation, particularly in systems involving particle aggregation, the dynamics of the population are influenced by two primary terms: the birth term and the death term. These terms are crucial for understanding how the population evolves over time, especially in processes like aggregation where particles combine to form larger entities.

Main Body

Birth Term

The birth term, denoted as $\overline{B_{j_i}}$, represents the contribution to the j-th moment of the population due to the formation of new particles. In aggregation processes, this typically involves smaller particles combining to form larger particles, which then contribute positively to the population moment. Key factors influencing the birth term include:

 Particle Size Distribution (PSD): The distribution of particle sizes at any given time, denoted as n = n(L, t), affects how

Ask me something..





Scalable GraphRAG-based Knowledge System for Domain Specific Workflow Automation with LLMs

Master thesis 3816

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Introduction

Current status of manufacturing industry

- Knowledge-Intensive: complex processes, scientific models
- Management of Knowledge

Traditional knowledge utilization method

- Storage of data in Database
- Large Language Model(LLM) Retrieval
 Challenge

How can LLM utilize the local knowledge?

Formula-Centric KnowledgeKnowledge:
Particle Size Balance ModelFormula Expression:
 $\frac{\partial n}{\partial t} + \frac{\partial (Gn)}{\partial L} = B - D$ Formula Explanation:
This equation is commonly used to describe the
dynamic behavior of a particle,...Explanation of Terms:
• B=B(L,t):The birth rate of the particle,...
• D=D(L,t):The death rate of the particle,......Knowledge Source:
Scientific paper

-Knowledge System

Knowledge Graph-based Retrieval + LLM Reasoning + QA Chatbot



Contents

- Introduction
- Background
- System design
- Evaluation & Analysis
- Conclusion & Outlook

• LLM Only

- Retrieval-Augmented Generation (RAG)
- Graph-based RAG(GraphRAG)



-Limitations:-

- No Access to Local Knowledge
- Risk of Data Leakage

-Question:

How can domain knowledge be incorporated?

Retrieval-Augmented Generation (RAG)

Enter your question (or type 'exit' to quit): in the population balance model, what factors are related to the birth term and the death term?

>>> Retrieving context (RAG retrieval)...

=== Retrieved Context ===

--- Retrieved Chunk 1 ---

population balance equation models to simulate secondary

nucleation processes have been derived. The first simulates

attrition as a breakage term, and growth rate is the result of size-dependent solubility. The second model considers attrition as a boundary condition at zero crystal size, where the expression for secondary nucleation rate already takes into account the effect of supersaturation, while the growth rate is size-independent. The two models are proven equivalent in the growth regime, thus

-- Retrieved Chunk 2 ---

function is n - dN/dL. Equation 1 relates the rate of change of number of particles in the differential size range L to L + dL, to the rates of growth into and out of that range and the rates of pirth and death in the size range.

In writing the population balance as Eq. 1, only growth, of the three phenomena of interest, appears explicitly. Before considering how appropriate birth and death functions might be formed for the other phenomena, a small simplification may be made to

--- Retrieved Chunk 3 ---

into either the "breakage model" or the "nucleation model". Model Equations. The model consists of the following population balance equation and solute mass balance:

Limitations:

- Chunk-based Context Loss
- Limited Relational Understanding





Background Literature Review



[1] RAG for knowledge-intensive NLP





[2] HybridRAG: Integrating Knowledge Graphs and Vector Retrieval



[5] Benchmarking the Impact of Knowledge Graphs on LLM SQL QA



[6] GraphRAG: Global Sensemaking with Graph-Based Retrieval

LLM only vs RAG vs GraphRAG

Criteria	LLM only	LLM + RAG	LLM + GraphRAG
Real-time Knowlege Access	Static, Pretrained	Dynamic, Unstructured	Dynamic, Structured
Domain Accuracy	Low (30%-60%) [3]	Improved (50%-80%) [1,2,3,6]	High (70%-90%) [2,4,5,6]
Explainable	Weak	Weak	Strong
System Complexity	Low	Moderate	High

- System Overview
- Retrieval Enhancement

System design System Overview



System design System Overview



System design System Overview



- System Overview
- Retrieval Enhancement
 - GraphRAG
 - + Chain of Thought
 - + Semantic Alignment
 - + Verification
 - + Task Decomposition

Retrieval Enhancement

GraphRAG + CoT + Semantic Alignment + Verification + Task Decomposition



Retrieval Enhancement

CoT Multi-condition Tasks Handling

GraphRAG + CoT + Semantic Alignment + Verification + Task Decomposition

What is **CoT(Chain of Thought)**?

Step-by-step reasoning process

Task:

What is the formula explanation and formula expression of particle size balance model?

CoT:

Step 1: Identify keywords from user task.

Step 2: Partial matching for each keyword.

Step 3: Identify conditions and purposes.

Step 4: ...

Retrieval Enhancement

GraphRAG + CoT + Semantic Alignment + Verification + Task Decomposition



Limitation

Col

- Semantic Drift
- Vocabulary Inconsistency

Multi-condition Tasks Handling

Task:

What is the **explanation** of "**Particle** distribution model"?

Answer:

The database does not contain relevant information.

Retrieval Enha

Retrieval Enhancement

GraphRAG + CoT + Semantic Alignment + Verification + Task Decomposition



Semantic Alignment -

What is the **explanation** of "**Particle**

The **Particle distribution model**

This equation is commonly used to

describe the dynamic behavior ...

Explanation of Formula:

Limitation Noisy output

distribution model"?

Answer:

Task:

Semantic Drift & Vocabulary Inconsistency Mitigation

Retrieval Enhancement

GraphRAG + CoT + Semantic Alignment + Verification + Task Decomposition



— Verification ——— Noisy output Mitigation

System design Retrieval Enhancement

GraphRAG + CoT + Semantic Alignment + Verification + Task Decomposition



Task:

In the particle size balance simulation process, what factors determine the birth item and what factors are related to the death item?

Sub Task 1:

"What factors determine the **birth item** in the **particle size balance simulation process**?"

Sub Task 2:

What factors are related to the **death** item in the particle size balance simulation process?"

Retrieval Enhancement

GraphRAG + CoT + Semantic Alignment + Verification + Task Decomposition



Task Decomposition –
 Complex Tasks Handling

- Evaluation of information retrieval capability
- Evaluation of response generation quality

Evaluation of information retrieval capability

Experiment Design

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Step 2: Getrictical SanAtiGissiand Truth generation

For each method and bestion, compute:

Uniqueness and Verifiability

Total Number of items retriev	
Vector-based semantic retrieval	Local Knowledge Utilization
Basic GraphRAG	Relationship-driven Reasoning
With Chain-of-Thought reasoning	Multi-condition Tasks Handling
+ Semantic Alignment	Semantic Drift Mitigation
+ Iterative query verification	Noisy output Mitigation
+ Task decomposition for multi-agent reasoning	Complex Tasks Handling
	Fotal Number of items retriev Vector-based semantic retrieval Basic GraphRAG With Chain-of-Thought reasoning + Semantic Alignment + Iterative query verification + Task decomposition for multi-agent

Number of items correctly retrieved

Evaluation of information retrieval capability

Experiment Result & Analysis

Strategies Metrics	RAG	GraphRAG	GraphRAG +COT	GraphRAG +COT+SA	GraphRAG +COT+SA +Veri	GraphRAG +COT+SA +Veri+TD
Hit Rate	6.20%	25.50%	20.08%	21.90%	21.90%	14.95%
Recall	60.88%	58.06%	77.73%.	78.62%	82.69%	91.17%
Inference time	1.27s	2.79s	3.45s	5.67s	6.07s	8.20s

- Missing key knowledge is riskier than retrieving too much
- Strategy complexity ↑, Recall ↑, Precision↓, Inference time↑

- Evaluation of information retrieval capability
- Evaluation of response generation quality

Evaluation of response generation quality

Experiment Design

- LLM-based Judgment (Deepseek)
- Scoring Criteria

Metrics	Description		
Accuracy Score(0-10)	Do the results match the user's expected answer?		
Overfetching / Underfetching(0-10)	Does the result return too much irrelevant info or miss key data?		
Readability(0-10)	Are the results understandable, structured, and informative?		

Evaluation of response generation quality

Experiment Result:

Strategies Metrics	RAG	GraphRAG	GraphRAG +COT	GraphRAG +COT+SA	GraphRAG +COT+SA +Veri	GraphRAG +COT+SA +Veri+TD
Accuracy Score	7.48	7.67	8.77	8.91	8.89	8.91
Overfetching / Underfetching	6.38	6.53	6.81	6.88	7.02	6.88
Readability	8.34	8.42	8.42	8.42	8.40	8.39

Conclusion & Outlook

Conclusion & Outlook

Overall

- Strong performance over RAG baseline
 - Graphrag: 91% recall vs RAG: 60%,
 - Improved precision and information completeness
- Limitations
 - 91% Recall, not 100% Coverage
 - Graph DB Construction Effort

Outlook

- Fusion with multimodal data
 - Text + Graph + Image , Cross-modal links
- Expansion into digital twin
 - Information Relationship Mapping
 - Digital twin environment



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Thank you!



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