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Journal Club: A review of large language models and autonomous agents in chemistry

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A review of large language models and autonomous agents in chemistry

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Large language models (LLMs) have emerged as powerful tools in chemistry, significantly impacting molecule design, property prediction, and synthesis optimization. This review highlights LLM capabilities in these domains and their potential to accelerate scientific discovery through automation. We also review LLM-based autonomous agents: LLMs with a broader set of tools to interact with their surrounding environment. These agents perform diverse tasks such as paper scraping, interfacing with automated laboratories, and synthesis planning. As agents are an emerging topic, we extend the scope of our review of agents beyond chemistry and discuss across any scientific domains. This review covers the recent history, current capabilities, and design of LLMs and autonomous agents, addressing specific challenges, opportunities, and future directions in chemistry. Key challenges include data quality and integration, model interpretability, and the need for standard benchmarks, while future directions point towards more sophisticated multi-modal agents and enhanced collaboration between agents and experimental methods. Due to the quick pace of this field, a repository has been built to keep track of the latest studies: <https://github.com/ur-whitelab/LLMs-in-science>.

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LLM vs. Traditional ML

1. Traditional Machine Learning (ML)

- Support Vector Machines (SVM)
- Random Forests (RF)

2. Neural Network Architectures in Deep Learning

- Recurrent Neural Networks (RNN) – for sequential data (e.g., text, reaction steps)
- Convolutional Neural Networks (CNN) – for grid-like data (e.g., spectra, electron density maps)
- Graph Neural Networks (GNN) – for molecular graphs (atoms as nodes, bonds as edges)
- Transformers – for long-range sequence modeling (e.g., SMILES, protein sequences)



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Large Language Models (LLMs)

LLM Types

1. Encoder–decoder models

- Bidirectional and Auto-Regressive Transformers (BART)

2. Encoder-only models

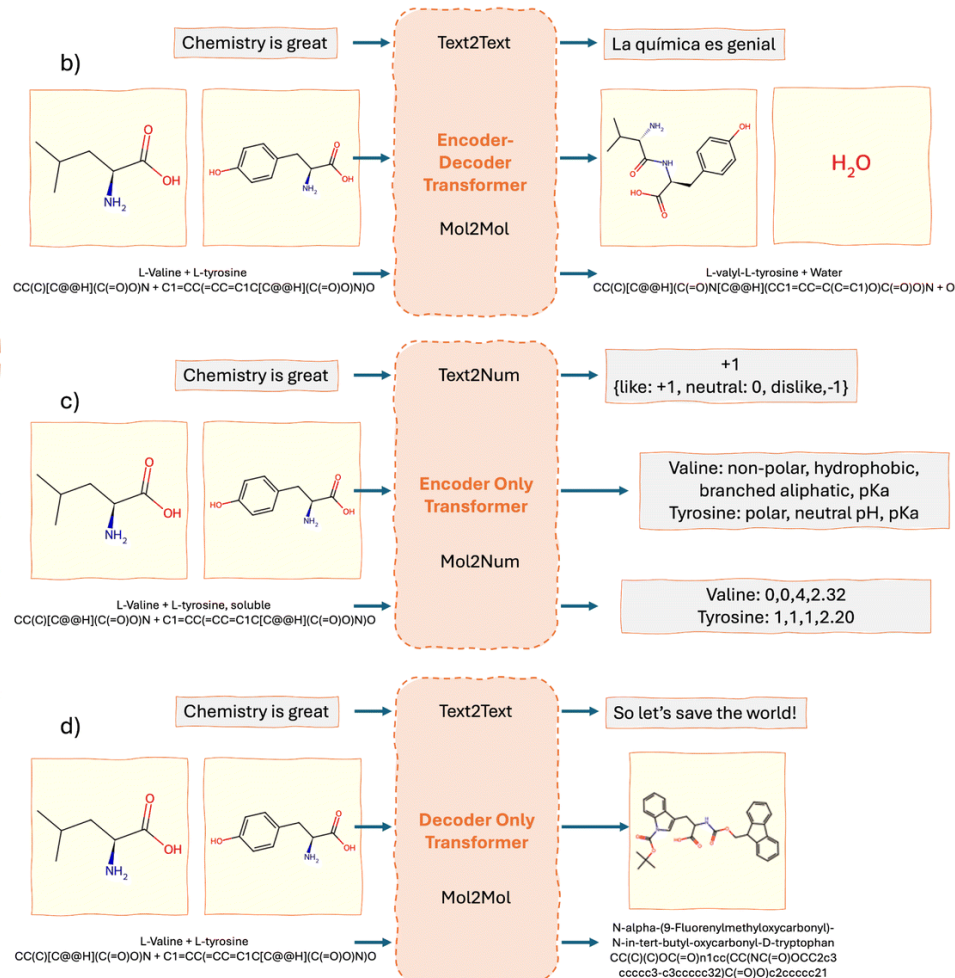
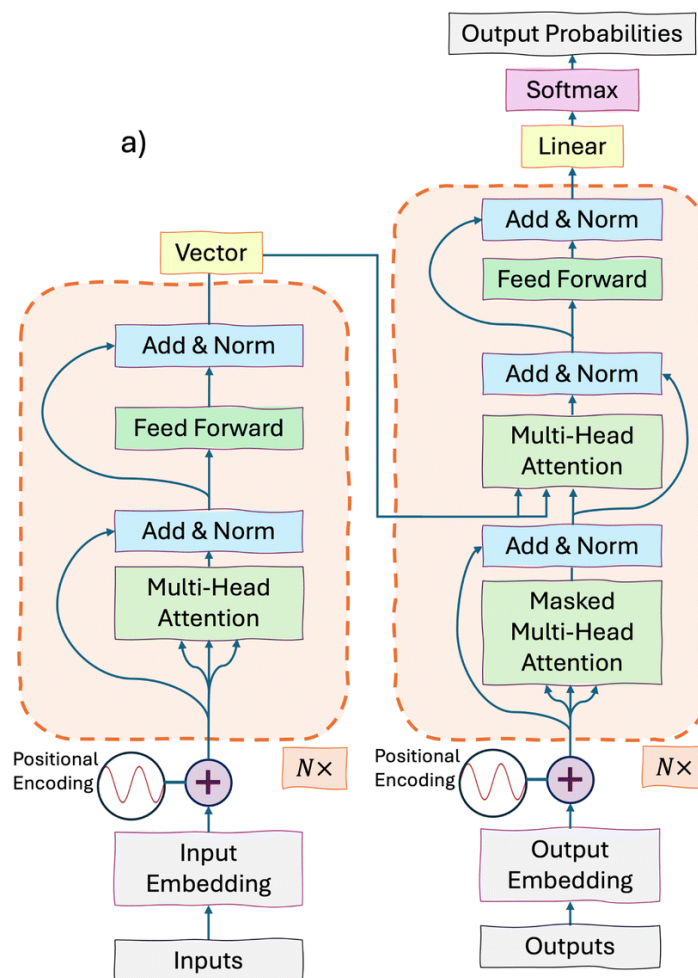
- Bidirectional Encoder Representations from Transformers (BERT)

3. Decoder-only models

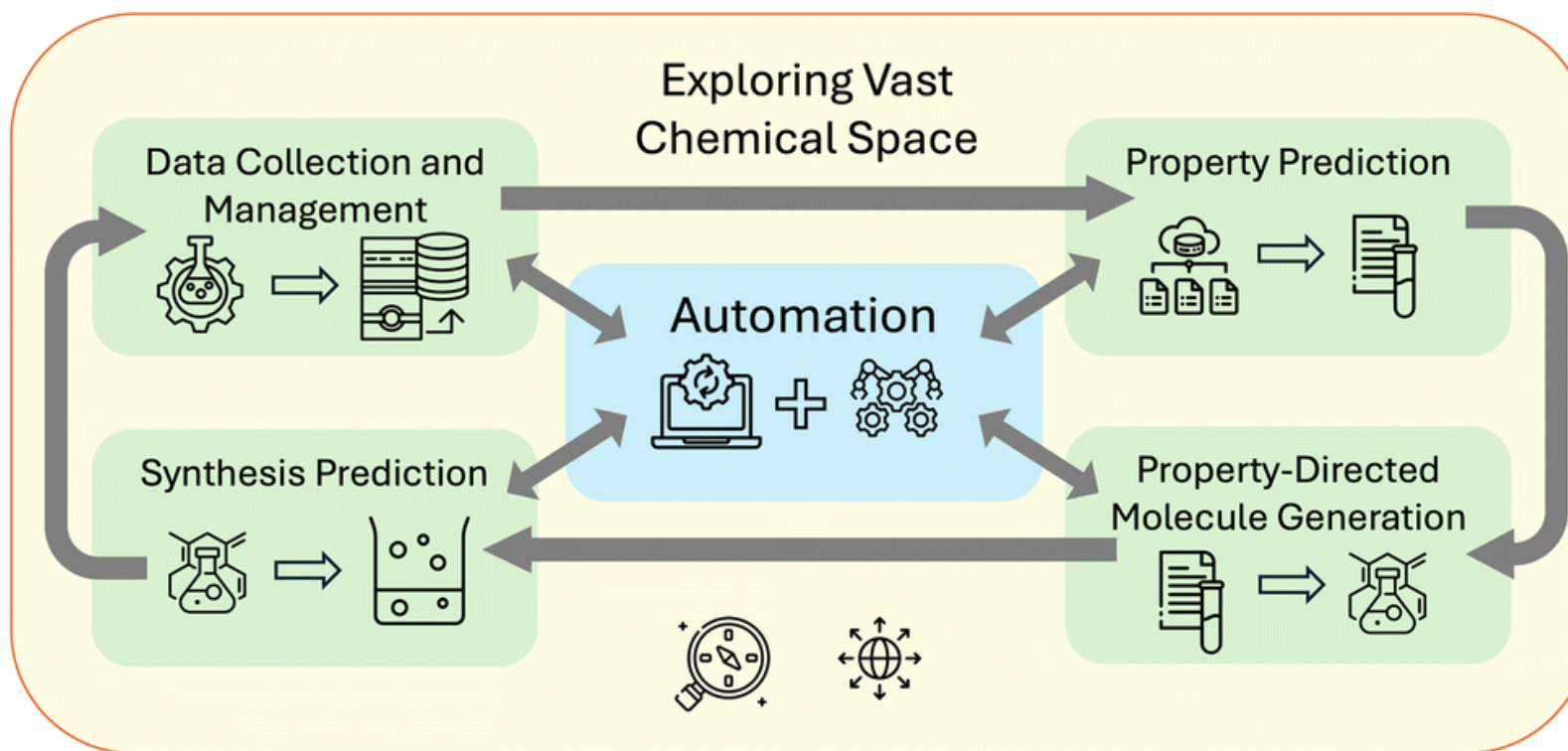
- Generative Pretrained Transformer (GPT)

4. Multi-task and multi-modal models

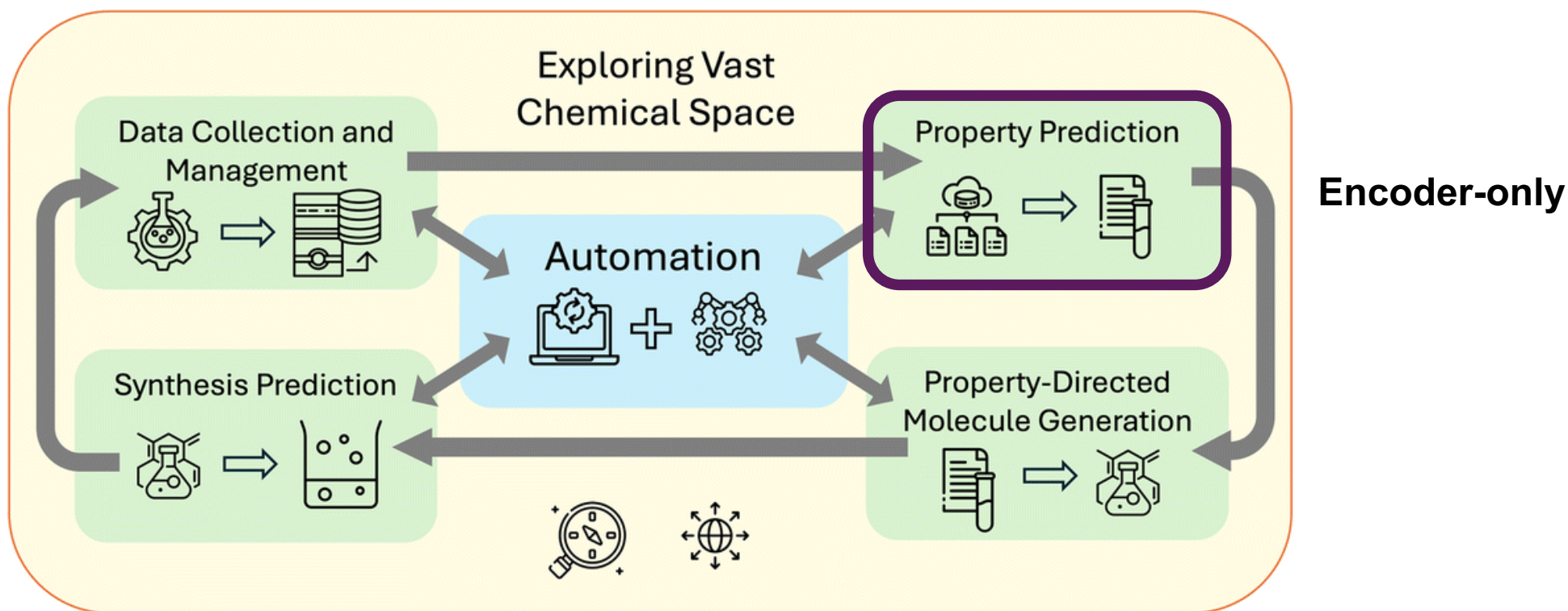
- Text-to-Text Transfer Transformer (T5)
- NExT-GPT



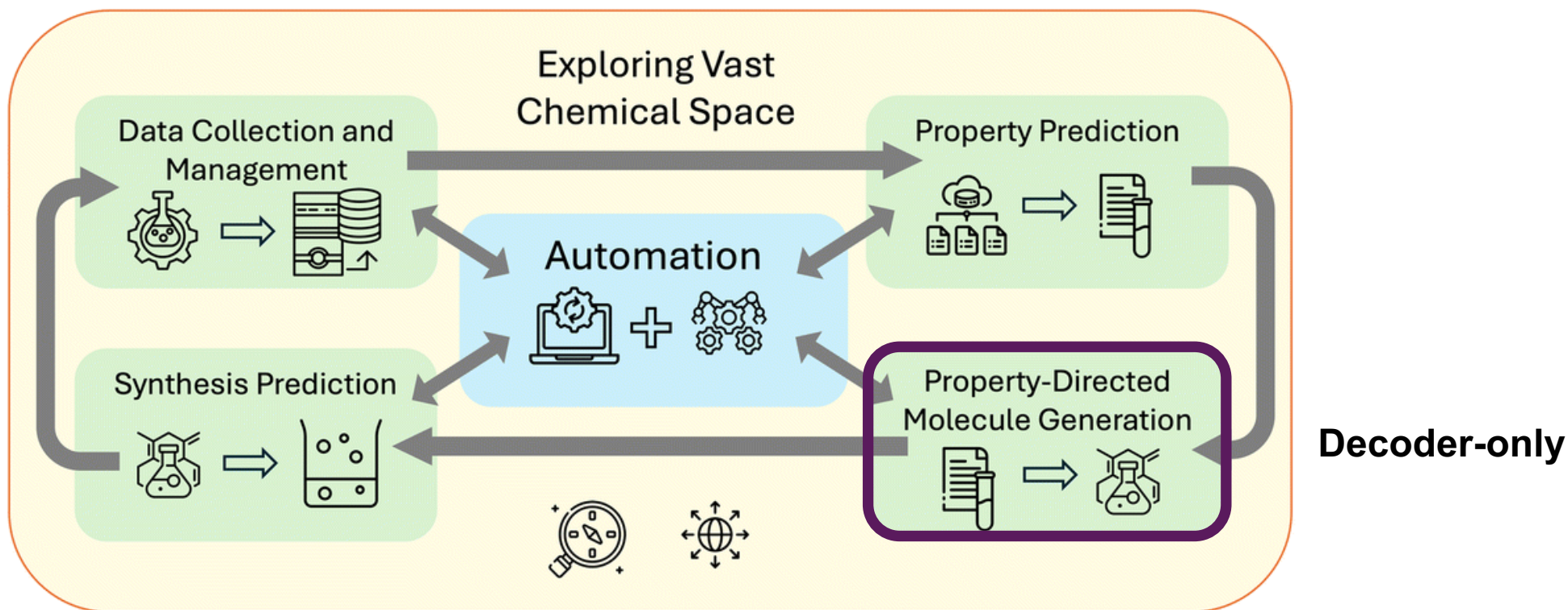
LLM in Chemistry



LLM in Chemistry

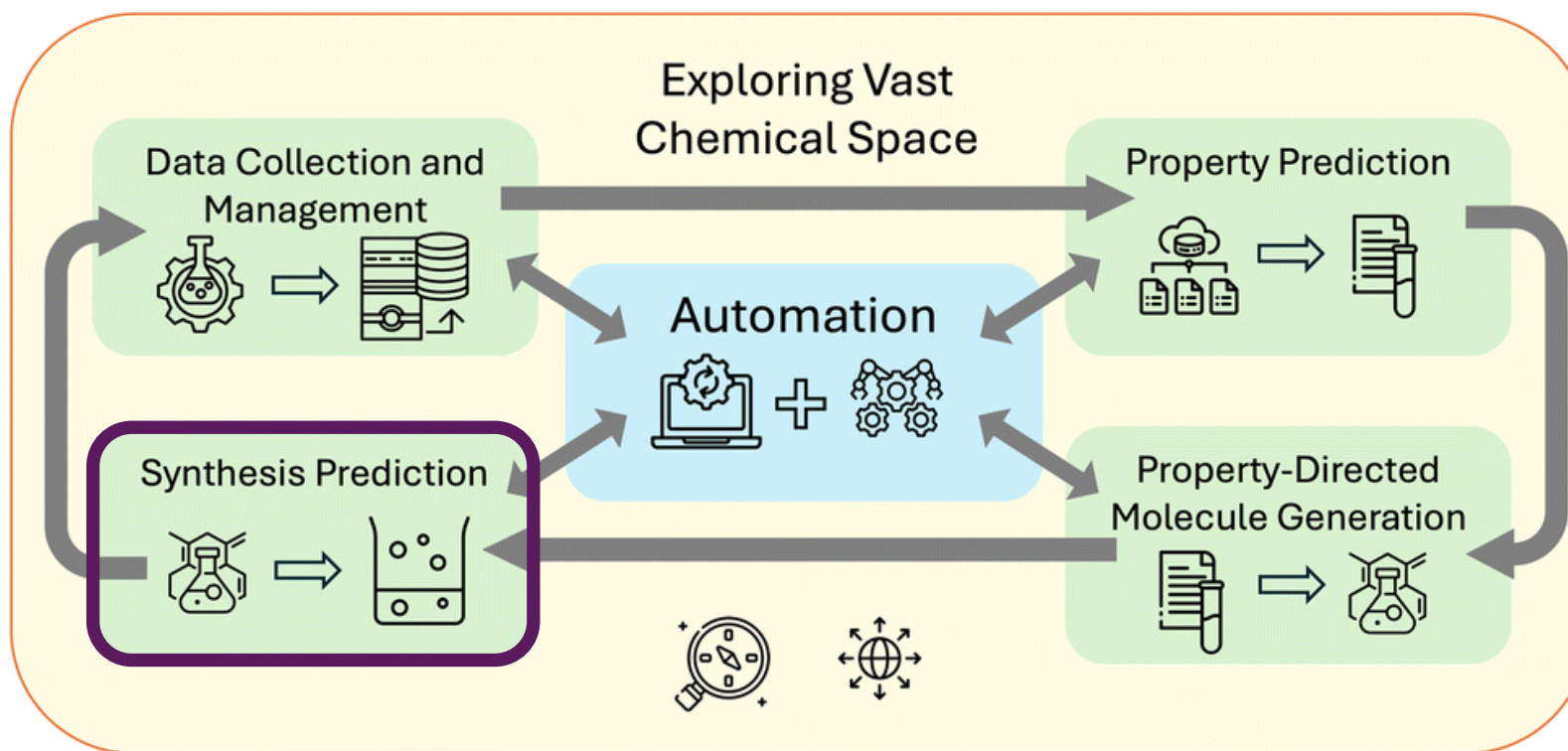


LLM in Chemistry

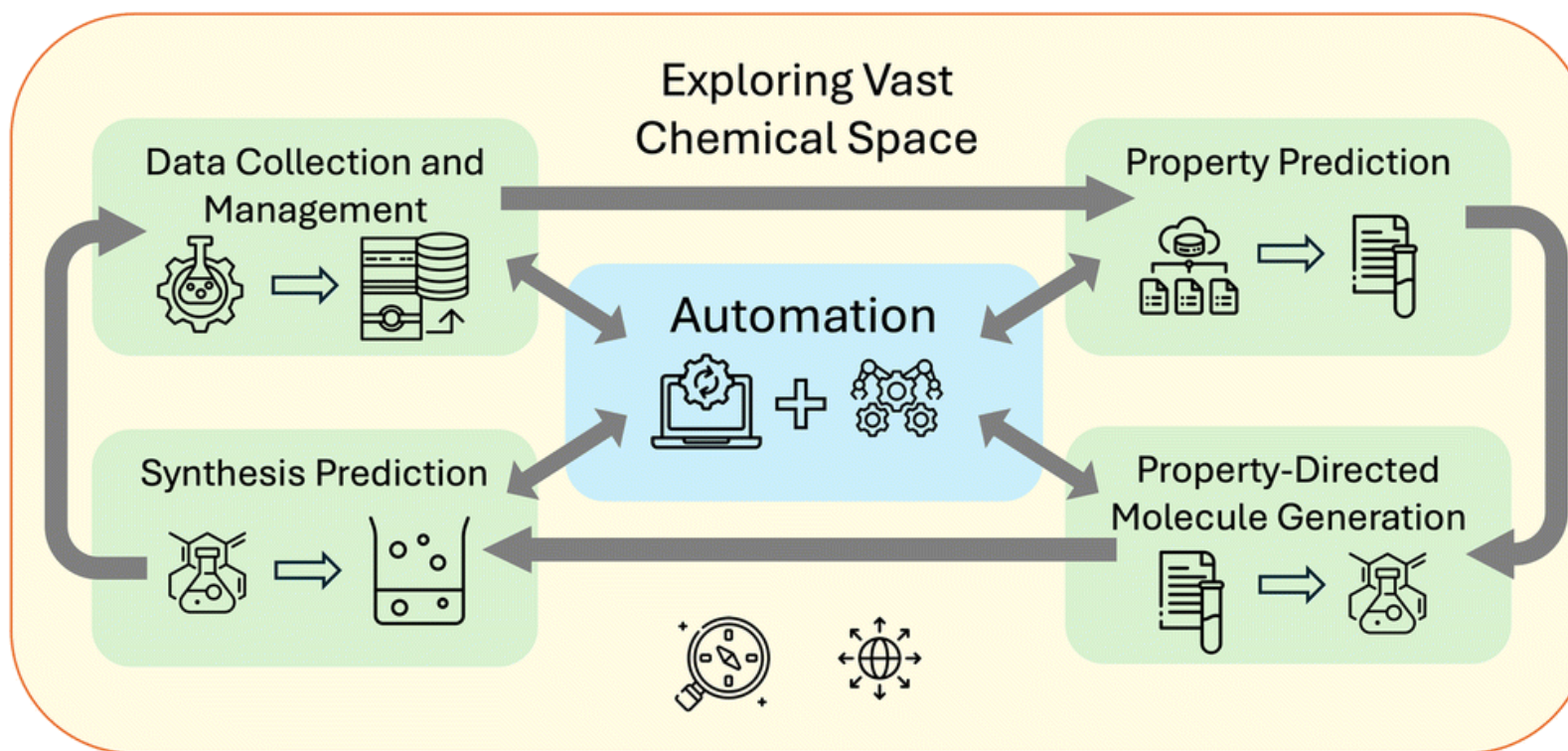


LLM in Chemistry

Encoder-decoder



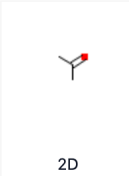

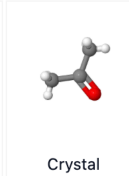


LLM in Chemistry



Mol-LLM: Molecular Representation Based on Strings

PubChem Acetone (Compound)

Acetone

PubChem CID	180
Structure	<div style="display: flex; justify-content: space-around; align-items: center;"> <div style="text-align: center;">  <p>2D</p> </div> <div style="text-align: center;">  <p>3D</p> </div> <div style="text-align: center;">  <p>Crystal</p> </div> </div>
Chemical Safety	<div style="display: flex; justify-content: center; gap: 10px;">   </div> <p style="text-align: center;">Flammable Irritant</p> <p style="text-align: center;">Laboratory Chemical Safety Summary (LCSS) Datasheet</p>
Molecular Formula	<p>C₃H₆O</p> <p>CH₃-CO-CH₃</p>
Synonyms	<p>acetone</p> <p>2-propanone</p> <p>67-64-1</p> <p>propanone</p> <p>Dimethyl ketone</p> <p>View More...</p>
Molecular Weight	<p>58.08 g/mol</p> <p><small>Computed by PubChem 2.2 (PubChem release 2021.10.14)</small></p>

PubChem Acetone (Compound)

2.1 Computed Descriptors ? ↗

2.1.1 IUPAC Name ? ↗

propan-2-one

Computed by Lexichem TK 2.7.0 (PubChem release 2021.10.14)

[▶ PubChem](#)

2.1.2 InChI ? ↗

InChI=1S/C3H6O/c1-3(2)4/h1-2H3

Computed by InChI 1.0.6 (PubChem release 2021.10.14)

[▶ PubChem](#)

2.1.3 InChIKey ? ↗

CSCPPACGZOO CGX-UHFFFAOYSA-N

Computed by InChI 1.0.6 (PubChem release 2021.10.14)

[▶ PubChem](#)

Simplified Molecular Input Line Entry System

2.1.4 SMILES ? ↗

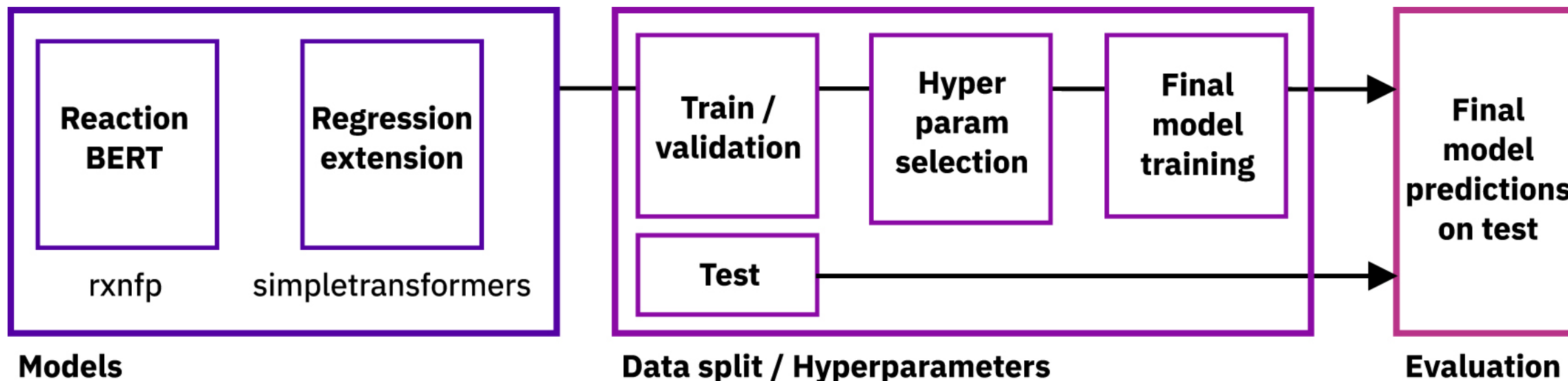
CC(=O)C

Computed by OEChem 2.3.0 (PubChem release 2024.12.12)

[▶ PubChem](#)



Predicts Reaction Yields Starting from Reaction SMILES



Brc1cccn1.CCOC(=O)c1cc(C)no1.CN1CCCN2CCCN=C12.COc1ccc(OC)c(P(C(C)(C)C(C)(C)C)c1-c1c(C(C)C)cc(C(C)C)cc1C(C)C.Cc1ccc(N)cc1.O=S(=O)(O[Pd]1c2ccccc2-c2ccccc2[NH2]1)C(F)(F)F>>Cc1ccc(Nc2cccn2)cc1

Reaction SMILES

Yield-BERT

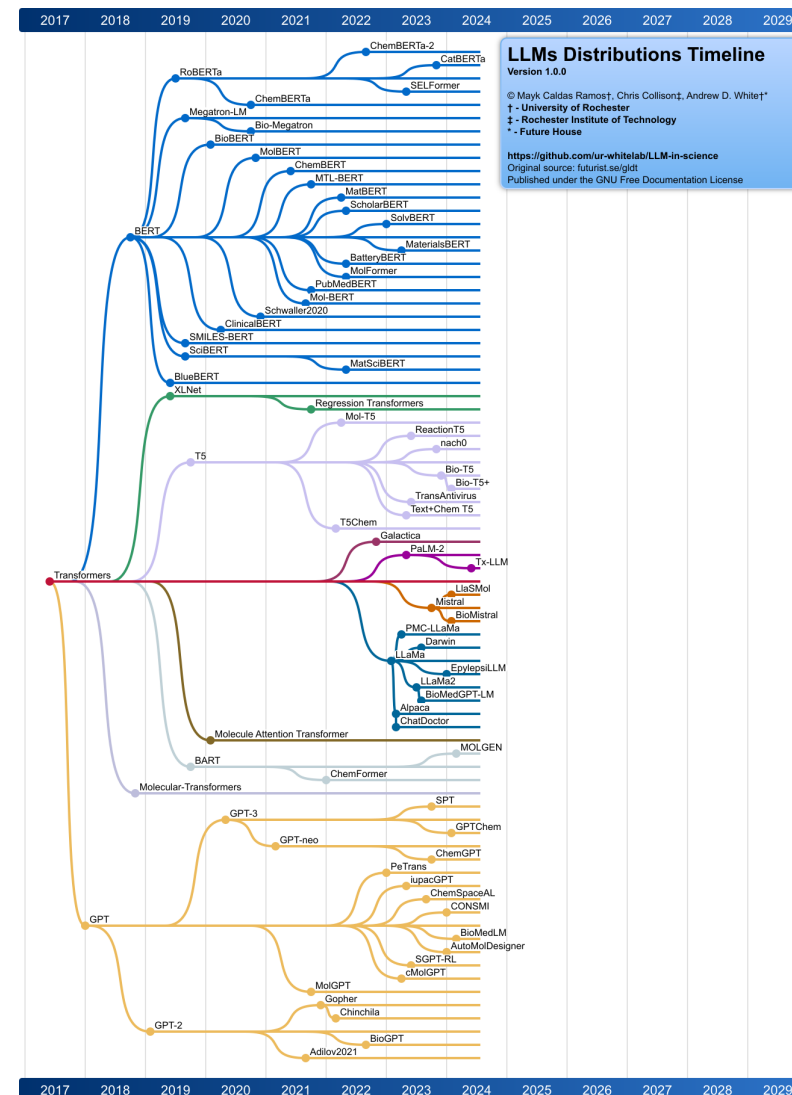
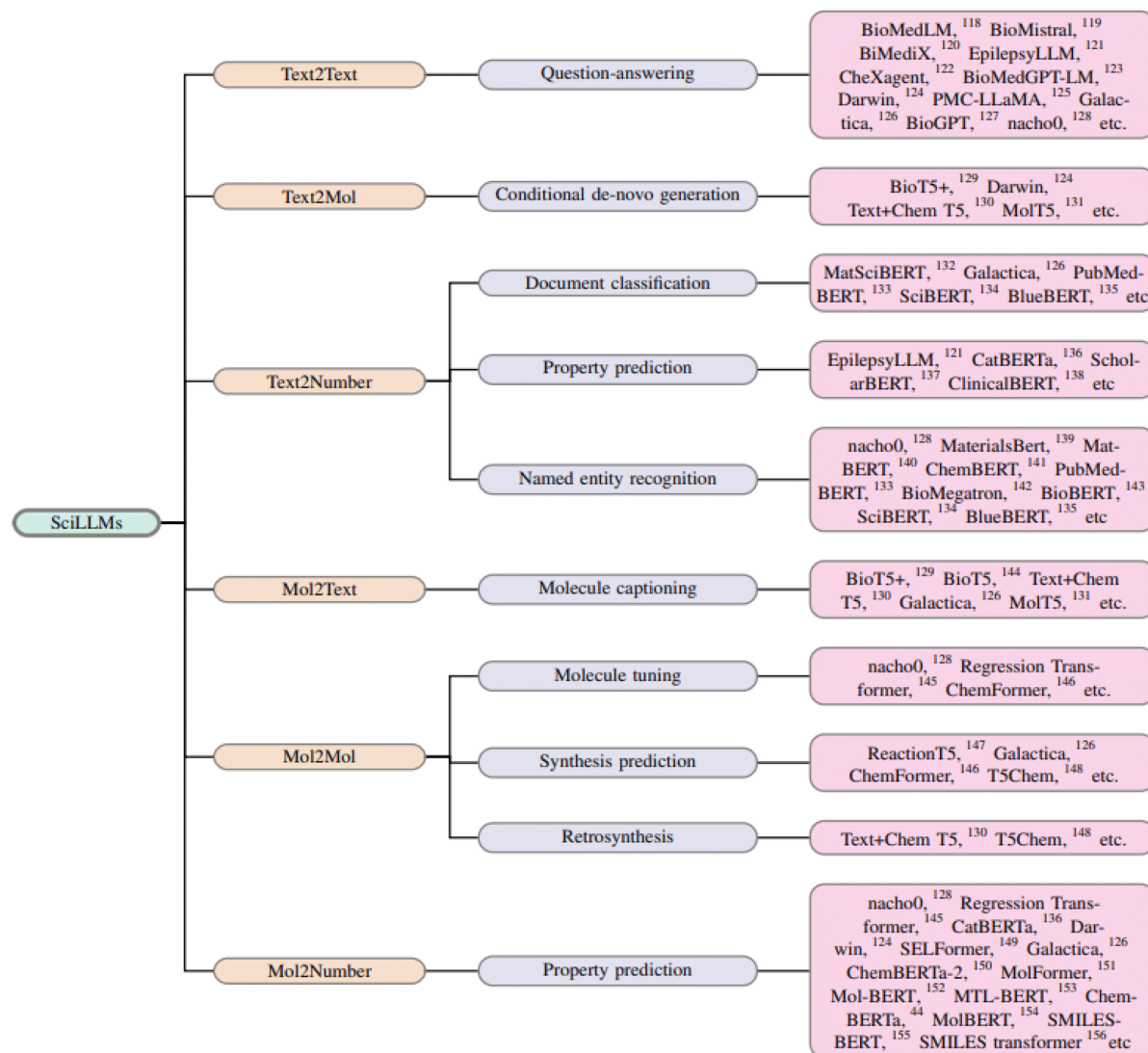


96.6%

Reaction yield

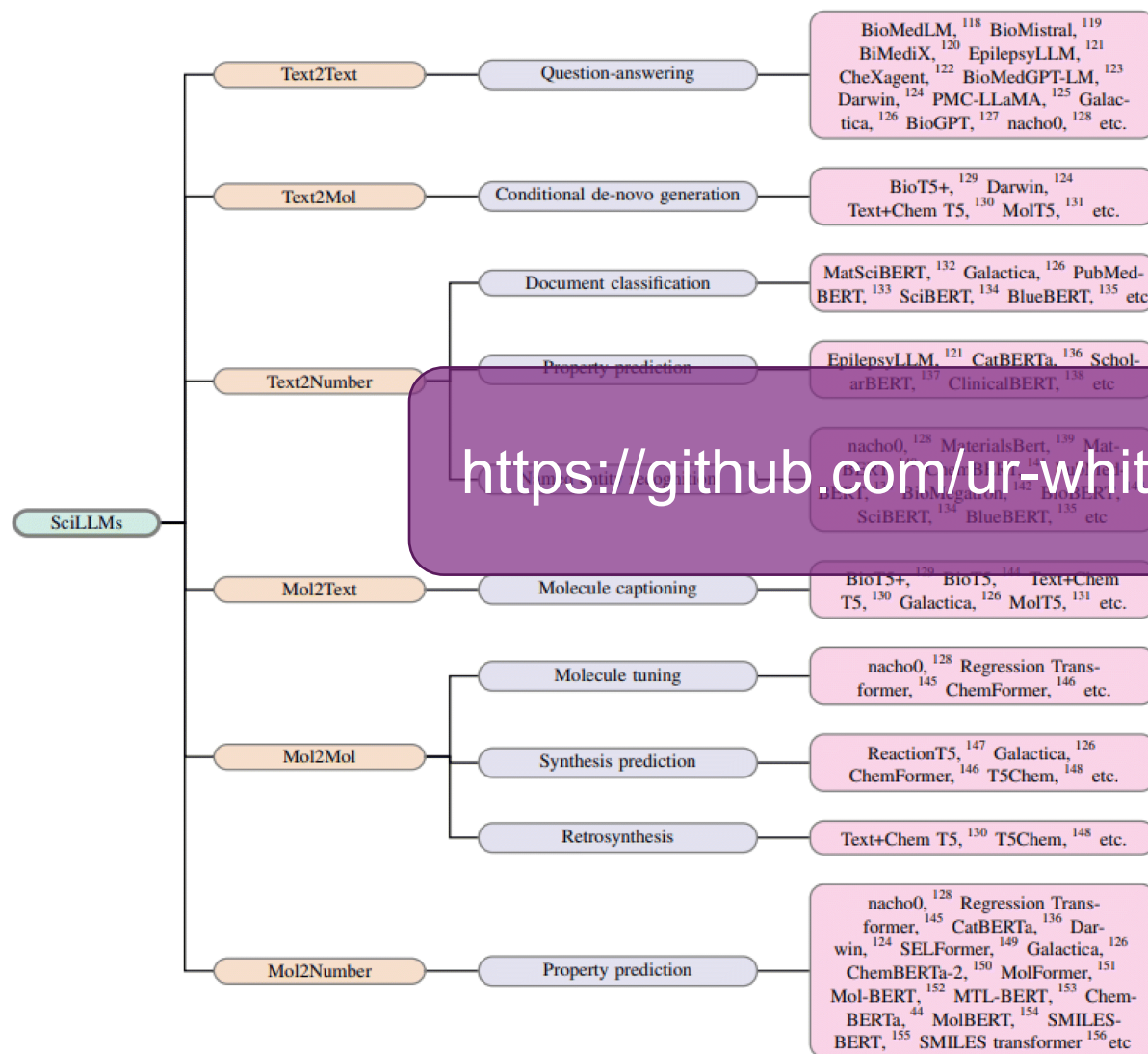


The Capabilities and Development of LLMs

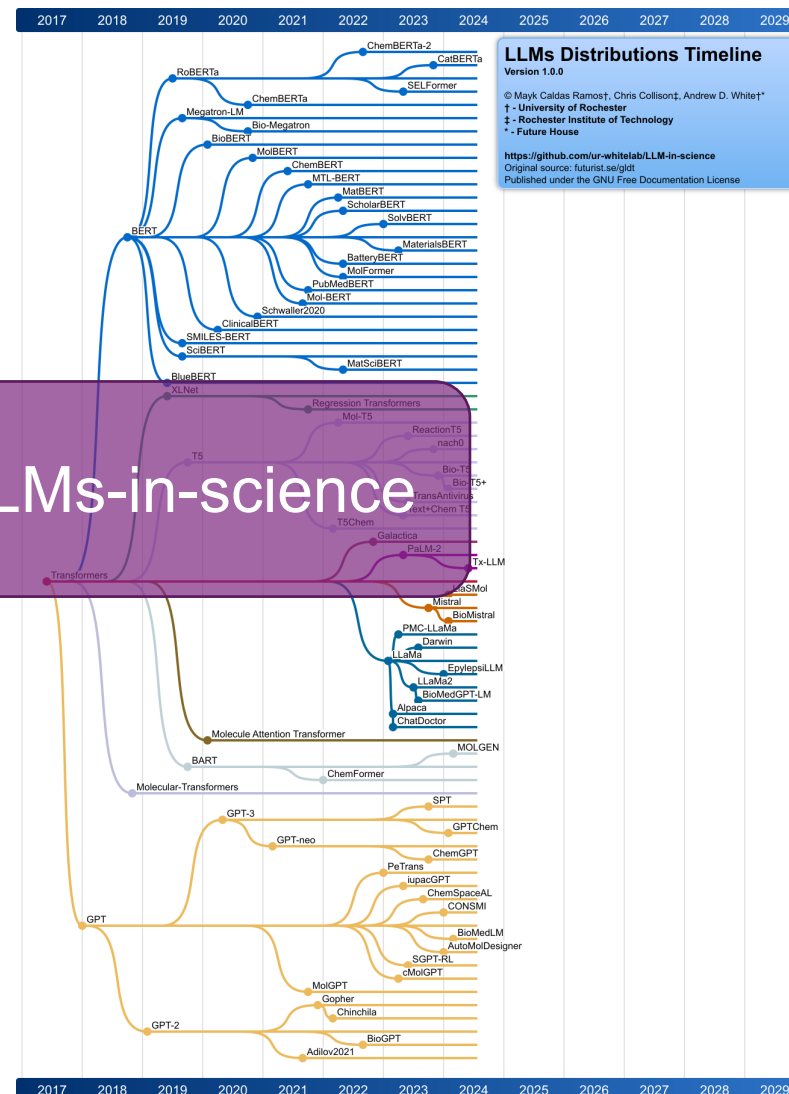




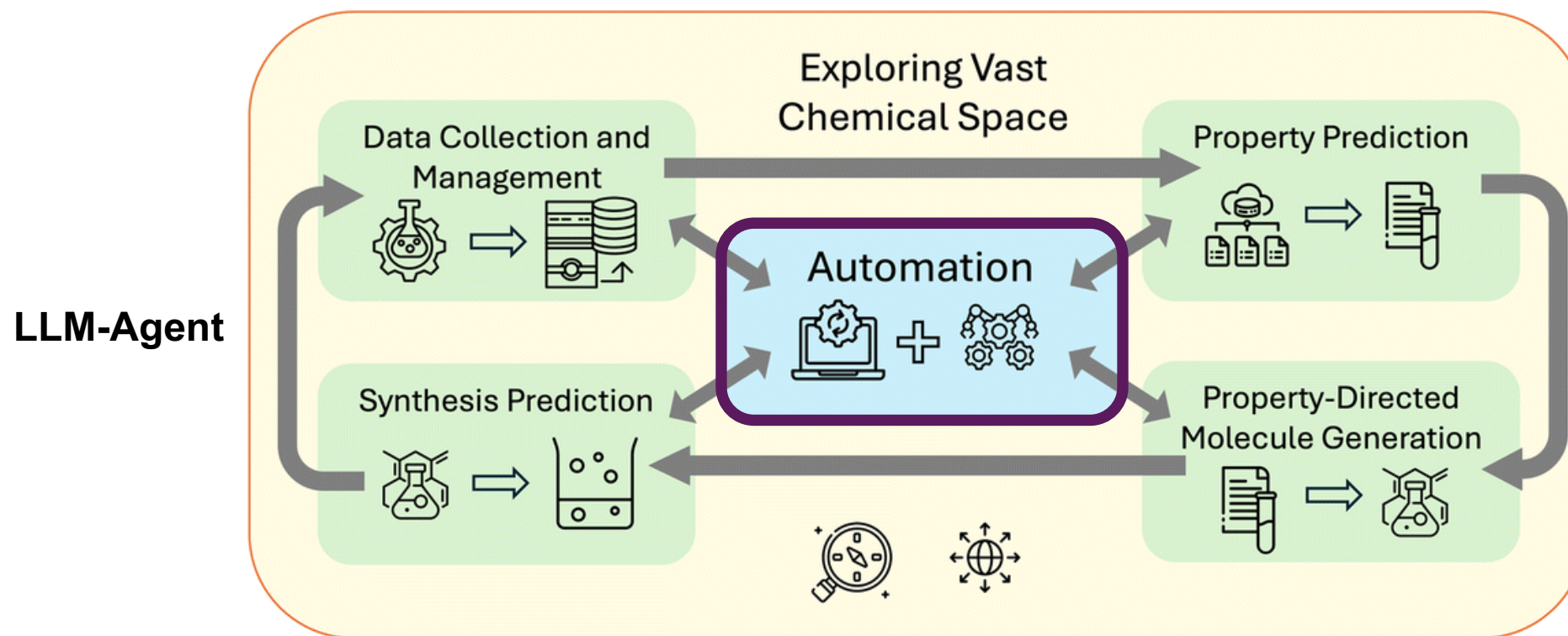
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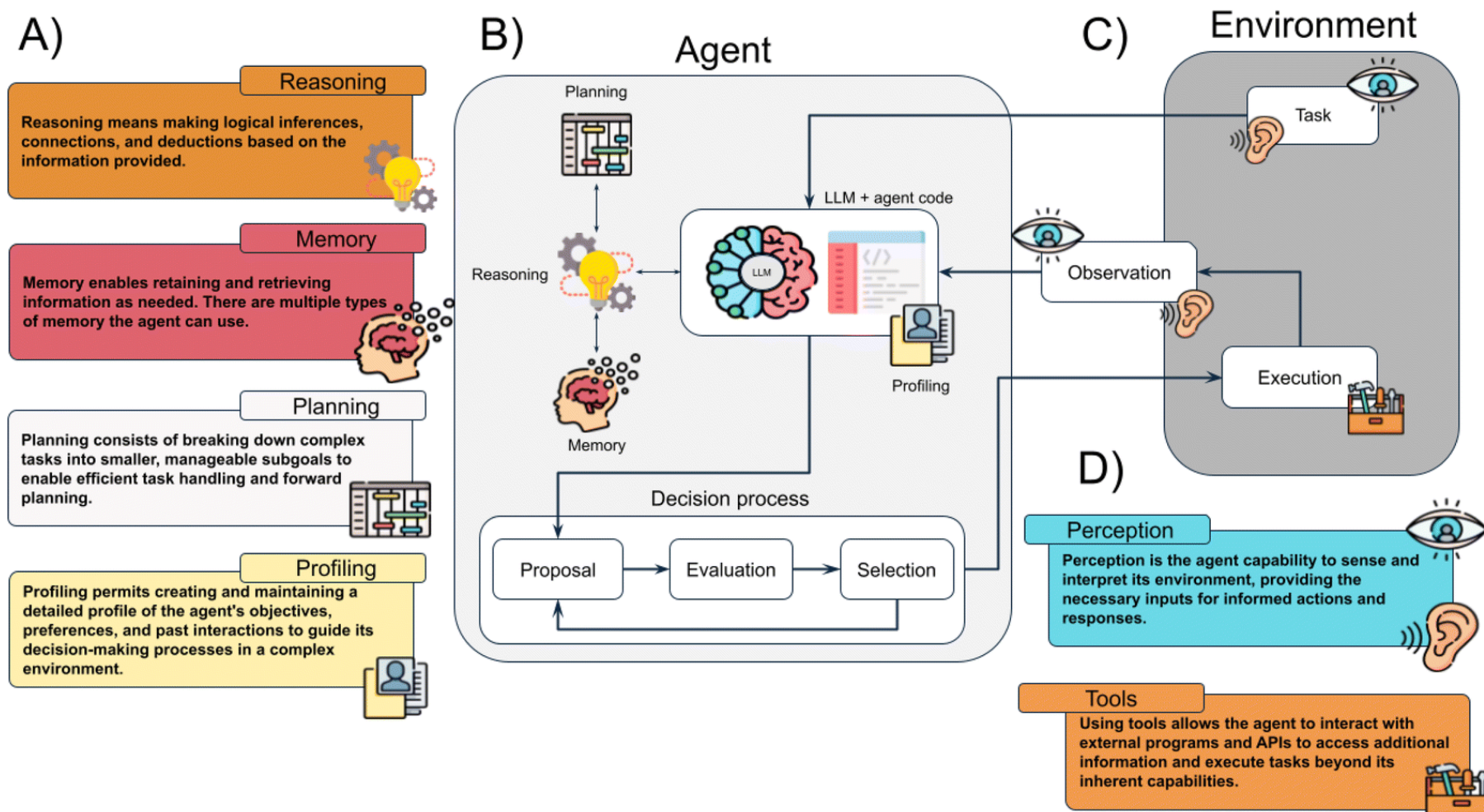
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Autonomous Agent



LLM-Agent





Limitations and Challenges

- 1. Lack of high-quality and diverse data**
- 2. Absence of standardized benchmarks**
- 3. Black-box nature and limited interpretability of LLMs**

Thank You