

# From Words to Wisdom: Automatically Generating Knowledge Graphs for Interpretable Educational Al



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#### LLMs Are Remarkable...

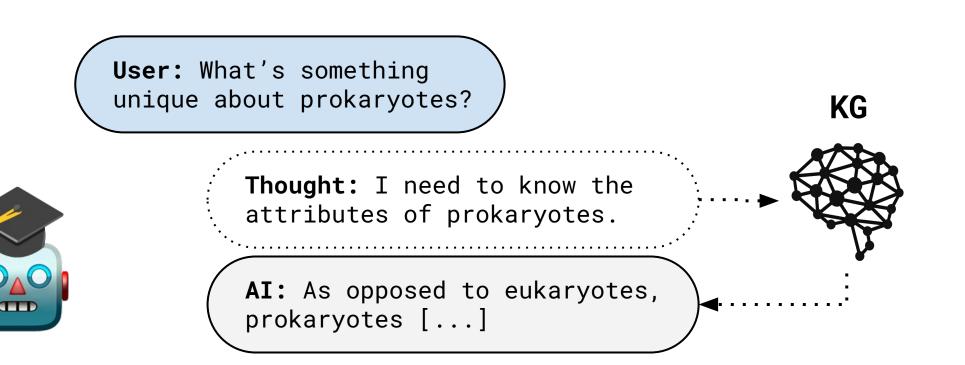
- ...remarkably prone to hallucination
- ...remarkably expensive to customize
- ...remarkably difficult to explain

How can a nonsense-generating, expensive, uninterpretable model be trusted in educational settings?

# Leveraging Knowledge Graphs

Knowledge injection: Augment LLMs with KGs to patch their weaknesses

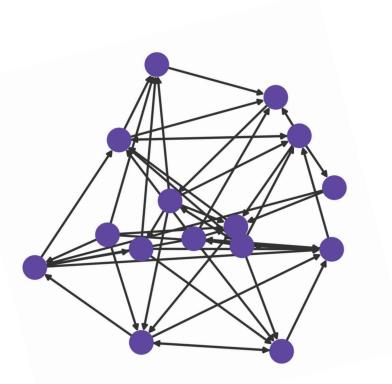
- Mitigate hallucination
- Improve interpretability & explainability
- ✓ Logical reasoning!



# **Knowledge Graphs 101**

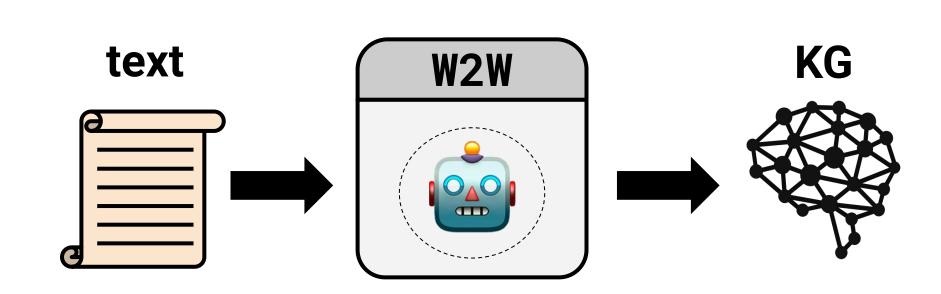
KGs are semantic structures that natively support **Q&A**, **reasoning**, etc.

- Nodes = entities
- Edges = relations
- Example: cell -(has) -> nucleus



### Words2Wisdom

Can LLMs effectively generate KGs from text? **Yes!** 



- KG preserves global relationships between ideas in text, despite chunking
- Contribution: Clause deconstruction
  - ✓ Parses complex & simple text uniformly
  - Aligns text to KG syntax to improve quality of KG components
- Quick, cost-effective at scale
- Domain-independent
- Intuitive user interface; command-line availability

### Demonstration

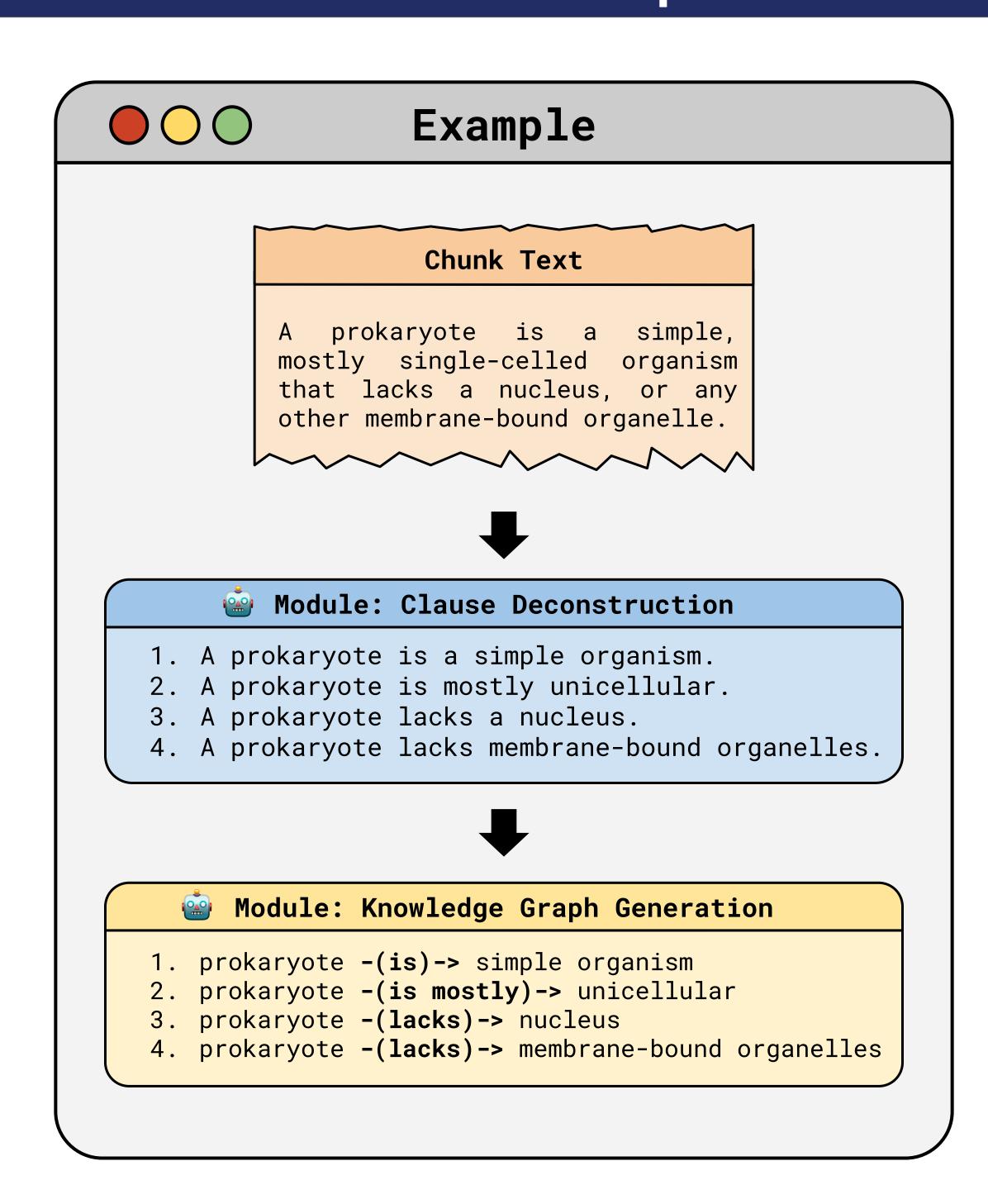
#### HF Space



#### **GitHub**



## Words2Wisdom: Pipeline



# **Evaluation (GPT-4)**

GPT-3.5 Pipeline	Component Integrity			Polovanco	Comprehensiveness
	subject	relation	object	- Nelevance	Comprehensiveness
KGG	0.93	0.92	0.82	0.91	0.74
CD + KGG	0.93	0.95	0.87	0.93	0.74

However, KG construction is traditionally resource-intensive...