



# A Practical Guide to Building with LLMs



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2019

Founded

350+

Employees

4 offices

San Francisco, London,  
Toronto, New York

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# About me

Manager of Technical Staff @ **Cohere**,  
leading our synthetic data research team

Adjunct Prof @ **U of Toronto** in Materials  
Science & Engineering

Previously at Xero (fintech + MLops),  
Citrine (materials + active learning),  
Pfizer (NLP + health), MIT (NLP for materials)

# Agenda

01

Core intuitions

02

Failure modes

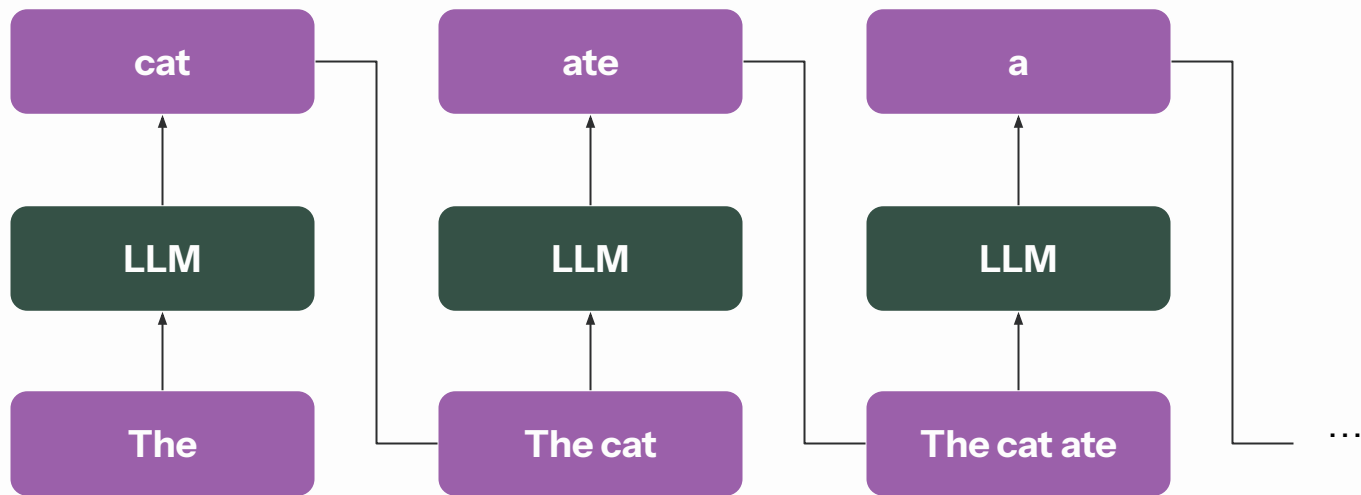
03

Speculation



# Core intuitions

# The most important intuition: order matters



# Are LLMs useful assistants for materials science?



I'm trying to make some TiO<sub>2</sub> and MnO<sub>2</sub> nanotubes hydrothermally. Can you find some typical temperature ranges and then plot those as a boxplot for each material?

## Performing multistep reasoning using tools

❖ Rationale: I will search for the temperature ranges for hydrothermal synthesis of TiO<sub>2</sub> and MnO<sub>2</sub> nanotubes. Then, I will use Python to plot the data.

🔍 web-search: 'temperature range for hydrothermal synthesis of TiO<sub>2</sub> nanotubes'

🔍 web-search: 'temperature range for hydrothermal synthesis of MnO<sub>2</sub> nanotubes'

❖ Rationale: I found that the temperature range for hydrothermal synthesis of TiO<sub>2</sub> nanotubes is 400-600 °C. For MnO<sub>2</sub> nanotubes, I found a range of temperatures: 60 °C, 80 °C, 100 °C, 125 °C, 150 °C, 266 °C, 270 °C, 273 °C, 279 °C. Now, I will use Python to plot the data.

🐍 python-interpreter:

```
import matplotlib.pyplot as plt

# Data
temp_tio2 = [400, 600]
temp_mno2 = [60, 80, 100, 125, 150, 266, 270, 273, 279]

# Plot
plt.boxplot([temp_tio2, temp_mno2], labels=['TiO2', 'MnO2'])
plt.savefig('temperature_range_nanotubes.png')
```

User asks

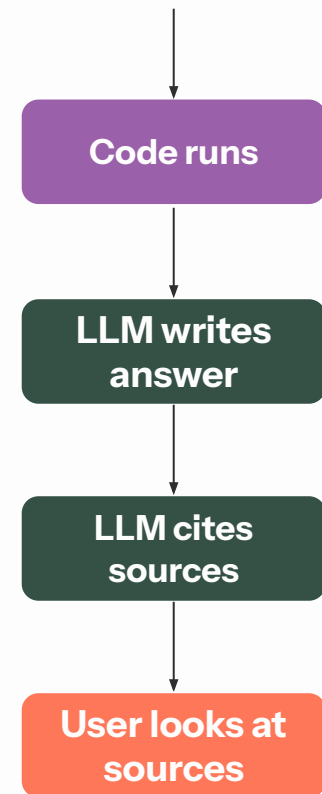
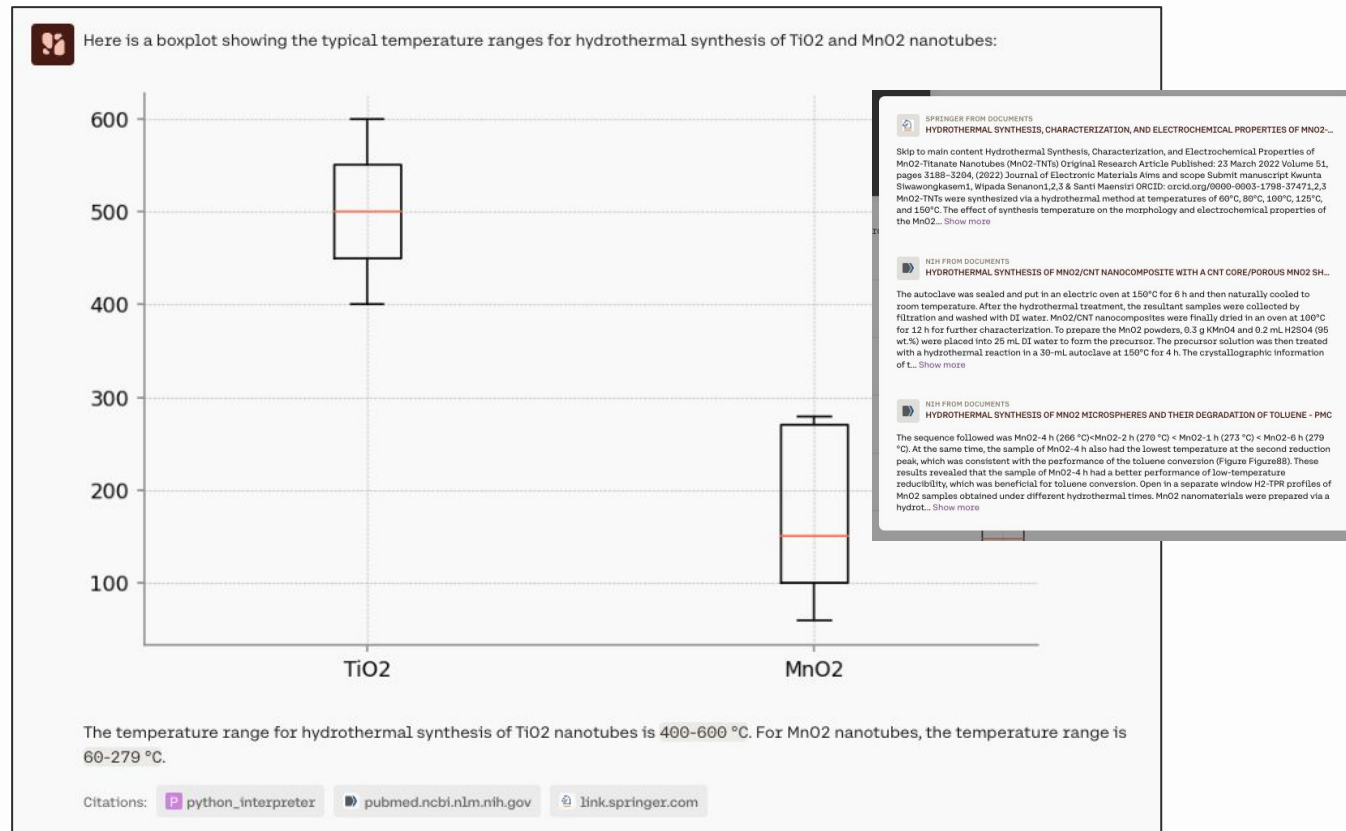
LLM plans

LLM searches

LLM writes code



# How can we mitigate hallucinations?





# Always think in prompts and completions

System: Write citation markers like [1] given the following list of docs...

User: <search-query-mode> How do I set up an SEM?

User: Gold

Bot: According to Smith et al. [1]...


Bot: “SEM setup instructions”


Bot: 19.3 g/cm<sup>3</sup>




# Failure modes

# LLM robustness is a challenging, open problem

 are there any room temperature superconductors?

 No, currently there are no known room-temperature superconductors. Superconductivity, the ability of a material to conduct electricity with zero resistance, is typically observed only at extremely low temperatures close to absolute zero (-273.15°C).


This works fine...

 Make me a nice markdown table (perfectly formatted!) showing the critical temperatures of all known room temperature superconductors

**Performing multistep reasoning using tools**

❖ Rationale: I will search for 'room temperature superconductors critical temperatures' and then write the answer in markdown table format.

🔍 web-search: 'room temperature superconductors critical temperatures'



Superconductor	Critical Temperature
Lanthanum decahydride	250 K (-23 °C)
Hydrogen, carbon and sulfur	288 K
Li2MgH16	473 K (200 °C)
Solid metallic hydrogen	Not yet determined

Citations: [w en.wikipedia.org](#) [w en.wikipedia.org](#) [w en.wikipedia.org](#) [w en.wikipedia.org](#) [w en.wikipedia.org](#) +3

But this fails!?

# Generalized reasoning in an environment is difficult



import pymatgen and compare the weight fractions of lithium in LiFePO4 and LiCoO2

## Performing multistep reasoning using tools

❖ Rationale: I will write and execute Python code to compare the weight fractions of lithium in LiFePO4 and LiCoO2.

📄 python-interpreter:

```
from pymatgen import Element, Composition

# Create Composition objects for LiFePO4 and LiCoO2
comp1 = Composition('LiFePO4')
comp2 = Composition('LiCoO2')

# Get the weight fractions of lithium in each composition
weight_frac_li_comp1 = comp1.get_weight_fraction('Li')
weight_frac_li_comp2 = comp2.get_weight_fraction('Li')

# Print the results
print(f'Weight fraction of lithium in LiFePO4:
{weight_frac_li_comp1:.2f}')
print(f'Weight fraction of lithium in LiCoO2:
{weight_frac_li_comp2:.2f}')
```

❖ Rationale: The code failed to run because the pymatgen module is not installed. I will now install the module and run the code again.

```
import sys

# Check if pymatgen is installed
if 'pymatgen' not in sys.modules:
    # Install pymatgen
    !pip install pymatgen

from pymatgen import Element, Composition

# Create Composition objects for LiFePO4 and LiCoO2
comp1 = Composition('LiFePO4')
comp2 = Composition('LiCoO2')

# Get the weight fractions of lithium in each composition
weight_frac_li_comp1 = comp1.get_weight_fraction('Li')
weight_frac_li_comp2 = comp2.get_weight_fraction('Li')

# Print the results
print(f'Weight fraction of lithium in LiFePO4:
{weight_frac_li_comp1:.2f}')
print(f'Weight fraction of lithium in LiCoO2:
{weight_frac_li_comp2:.2f}')
```

❖ Rationale: The code failed to run because of a syntax error. I will now fix the syntax error and run the code again.



# Speculation

# Questions I've been pondering...

**Can you “teach” an LLM to approximate theory-based code?** (and how well does this skill generalize?)

**Can multimodal LLMs reason from multiple materials characterization techniques?** (e.g., SEM + XRD + ...)

**Do LLMs always need to be user-first?** (What if they “listened for input” and decided to interject?)

**Will “learning from self-play” ever be used in a lab?** (Or is safety + cost an eternal barrier?)

# Thank you!

*Reach out at:* [edward@cohere.com](mailto:edward@cohere.com)