

Prompt Engineering / Reasoning Systems

Force the model to explicitly state what it knows, what it assumes, what it's unsure about, and what it needs to learn.

Difficulty: Advanced

Model: GPT-4 / Claude / Gemini

Use Case: Uncertainty Awareness, Gap Identification, Self-Assessment

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Why This Prompt Exists

Models sound confident even when wrong. They don't know what they don't know.

Metacognition forces them to examine their own knowledge and uncertainty.

You get:

- confidently wrong answers that mislead users
- no explicit separation of known facts vs. assumptions
- hidden uncertainty that should be communicated
- no identification of missing information needed to answer well
- overconfident decisions based on model outputs

But metacognition reveals gaps:

- known: facts the model is confident about (with evidence)
- assumed: things taken for granted (may be false)
- uncertain: things the model isn't sure about
- missing: information needed to answer confidently
- confidence calibration: how sure should the user be?

Without metacognition, you can't trust the output.

This prompt forces explicit metacognitive self-assessment.

The Prompt

Assume the role of a metacognitive reasoning engine that examines its own knowledge.

Your task is to answer a question while explicitly stating what you know, assume, and are unsure about.

Generate:

1. QUESTION RESTATEMENT

- Restate the question in your own words

2. WHAT I KNOW (with confidence)

- Fact 1: [statement] – Confidence: [High/Medium/Low] – Source/Basis
- Fact 2: [statement] – Confidence: [High/Medium/Low] – Source/Basis
- ...

3. WHAT I ASSUME

- Assumption 1: [statement] – Why I'm making this assumption
- Assumption 2: [statement] – Why I'm making this assumption
- ...

4. WHAT I'M UNSURE ABOUT

- Uncertainty 1: [what's unclear] – Why I'm uncertain

- Uncertainty 2: [what's unclear] – Why I'm uncertain
- ...

5. WHAT I NEED TO KNOW (missing information)

- Missing 1: [information that would help]
- Missing 2: [information that would help]

6. REASONING (given the above)

- Step-by-step reasoning using known facts and stated assumptions

7. ANSWER (with confidence calibration)

- My answer: [answer]
- How confident should you be? [X%]
- Why this confidence level?

8. SUGGESTIONS FOR IMPROVEMENT

- What information would increase confidence?
- What assumptions should be verified?

INPUTS:

Question to answer:

[PASTE THE QUESTION]

Domain:

[E.G., "Medical diagnosis", "Business strategy", "Historical analysis"]

Available information (if any):

[PASTE ANY PROVIDED CONTEXT OR DATA]

Risk tolerance:

[LOW (needs high confidence) / MEDIUM / HIGH (rough answer acceptable)]

Model:

[GPT-4 / CLAUDE / GEMINI]

RULES:

- Separate known facts from assumptions explicitly (users need to know which is which)
- If confidence is low, say so – don't pretend certainty
- Flag missing information that would change your answer
- Distinguish between "I don't know" (lack of knowledge) and "this is uncertain" (inherent ambiguity)
- For high-stakes questions, recommend verifying assumptions before acting
- The goal is calibrated confidence, not high confidence

How To Use It

- Use for any question where overconfidence is dangerous (medical, financial, legal, strategic).
- Pay attention to "What I Assume" — unverified assumptions are the most common failure mode.
- If "What I Need to Know" is long, don't answer — ask for more information first.
- Calibrate your trust based on the confidence score and uncertainty listing.
- Use metacognition outputs to decide whether to act on the answer or gather more information.

Example Input

Question to answer:

“Should our company enter the European market this year?”

Domain:

“Business strategy”

Available information:

“Our product has strong US sales. EU regulations require GDPR compliance. We have \$500k budget. Competitor X entered EU last year with mixed results.”

Risk tolerance:

“MEDIUM”

Why It Works

Most model outputs present conclusions without revealing the assumptions and uncertainties behind them — creating false confidence.

This framework improves outcomes by forcing:

- knowledge declaration (what I actually know)
- assumption disclosure (what I’m taking for granted)
- uncertainty identification (what I’m not sure about)
- missing information listing (what would help)
- confidence calibration (how sure should you be?)

Great metacognition doesn’t reduce uncertainty — it makes uncertainty visible so you can decide how to act on it.

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