

## Prompt Engineering / Chain-of-Thought

Find the shortest valid reasoning chain for a problem — reduces tokens while maintaining accuracy.

Difficulty: Advanced

Model: GPT-4 / Claude / Gemini

Use Case: Cost Optimization, Latency Reduction, Efficient Reasoning

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Why This Prompt Exists

Chain-of-thought is powerful but expensive. Many reasoning chains are longer than necessary — adding tokens, latency, and cost without improving accuracy.

You get:

- reasoning chains with redundant or unnecessary steps
- paying for tokens that don't improve the answer
- slower responses from overly verbose reasoning
- context window filled with fluff instead of useful reasoning
- no systematic way to find the minimal effective chain

But minimal chains have properties:

- necessary steps: cannot be removed without breaking correctness
- redundant steps: can be removed with no impact
- compressible steps: can be stated more concisely
- parallelizable steps: can be reordered or combined
- implicit steps: don't need to be stated (model can infer)

Without optimization, you waste tokens.

This prompt finds the minimal reasoning chain that still produces correct answers.

The Prompt

Assume the role of a reasoning efficiency engineer who compresses reasoning chains.

Your task is to find the minimal set of steps needed to solve a problem correctly.

Generate:

1. ORIGINAL REASONING CHAIN
  - All steps as originally written

## 2. STEP NECESSITY ANALYSIS

Step	Content	Necessary? (Yes/No)	Why
1	[step]	Yes/No	[reason]
2	[step]	Yes/No	[reason]
...	...	...	...

## 3. REDUNDANT STEPS (can be removed)

- List with rationale

## 4. COMPRESSIBLE STEPS (can be shortened)

- Original vs. compressed version

## 5. MINIMAL REASONING CHAIN

- Only necessary steps, each stated concisely

## 6. VALIDATION

- Does the minimal chain still produce the correct answer? (Yes/No – test required)
- Token savings: [X] tokens saved ([Y]% reduction)
- Accuracy risk: (Low / Medium / High)

## 7. DEPLOYMENT RECOMMENDATION

- Use minimal chain for production (if low risk)
- Keep original for complex cases
- Test on edge cases before deploying

## INPUTS:

Original reasoning chain (full):

[PASTE THE REASONING WITH STEPS]

Problem type:

[ARITHMETIC / LOGIC / COMMON SENSE / PLANNING / OTHER]

Correct answer (for validation):

[THE KNOWN CORRECT ANSWER]

Model:

[GPT-4 / CLAUDE / GEMINI]

## RULES:

- A step is necessary if removing it changes the answer or breaks the reasoning
- A step is redundant if the model can infer it without explicit statement
- Compressed steps must preserve meaning (no shortcuts that change the logic)
- Test the minimal chain before deploying (some necessary steps are not obvious)
- Flag if minimal chain reduces accuracy for edge cases

#### How To Use It

- Run this on reasoning chains that will be used at high volume — token savings add up fast.
- Test the minimal chain on at least 10 edge cases before deploying.
- Don't over-optimize for low-stakes tasks (saving tokens isn't worth accuracy risk).
- Save the minimal chain as a template for similar problems.
- Re-optimize when you switch to a new model (different models need different steps).

#### Example Input

##### **Original reasoning chain:**

“Step 1: The problem asks for 15% of 280. Step 2: 15% means 15 per 100. Step 3: So 15% of 280 equals  $(15/100) \times 280$ . Step 4:  $15/100$  simplifies to 0.15. Step 5:  $0.15 \times 280 = (0.15 \times 200) + (0.15 \times 80)$ . Step 6:  $0.15 \times 200 = 30$ . Step 7:  $0.15 \times 80 = 12$ . Step 8:  $30 + 12 = 42$ . Step 9: Therefore, 15% of 280 is 42.”

##### **Problem type:**

“ARITHMETIC”

##### **Correct answer:**

“42”

### Why It Works

Most reasoning chains are written for clarity — which is good for learning — but inefficient for production.

This framework improves outcomes by forcing:

- step necessity analysis (what actually matters)
- redundancy identification (what can be removed)
- step compression (what can be shortened)
- validation testing (does it still work?)
- token savings calculation (quantified benefit)

Great minimal step optimization doesn't sacrifice accuracy — it removes waste while preserving correctness.

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