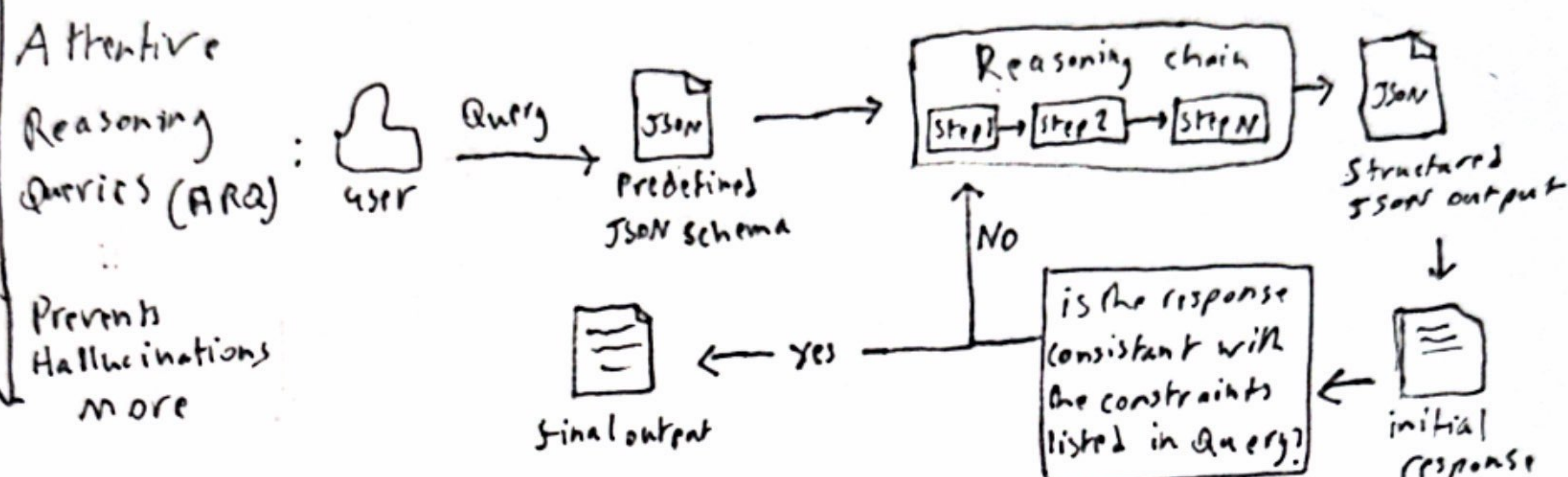
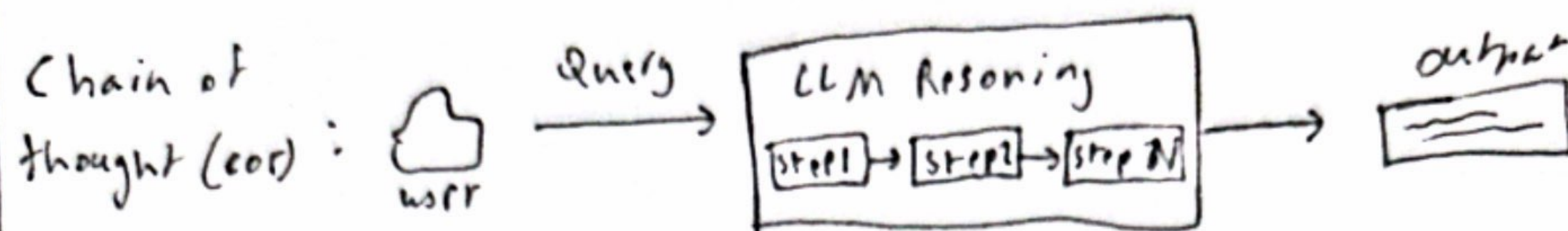
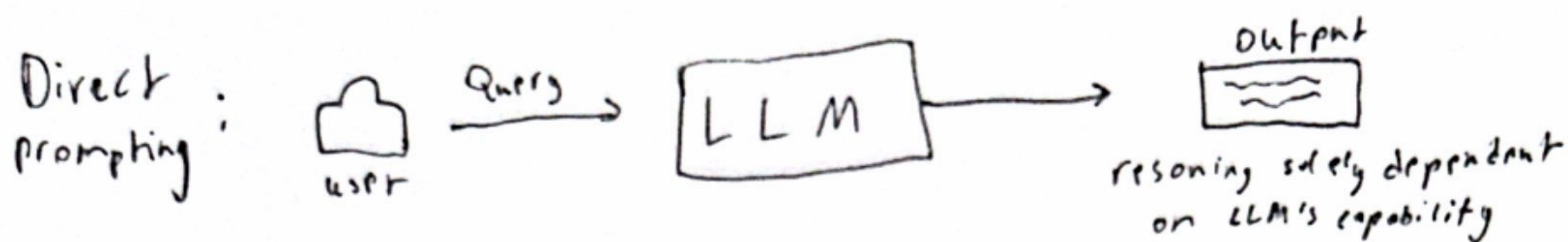


# More Prompt Engineering Approaches (simple, CoT, ARQ)



## Positional Embeddings

- in short: a positional embedding vector is a vector added to token embeddings to encode the position of each token in a sequence, so the model knows the order of the words

- Problem: Transformers like GPT process all tokens in parallel (unlike RNN) they don't have a idea of sequence order. so feeding it a sentence like: "The cat sat on the mat", without any extra info, they only see a bag of embeddings not which word came first.

- Solution: Positional embedding, a positional embedding is a vector that encodes the position of a token only (where in the sequence it is) position 0, 1, 2 etc, these are added to the token embedding (also vector) before feeding them into the transformer

So for input: ["the", "cat", "sat", ...]  $\xrightarrow[\text{learned from vocabulary}]{\text{Token embeddings}}$   $E(\text{"the"}), E(\text{"cat"}), E(\text{"sat"}), \dots$

$\xrightarrow[\text{Fixed or learned vectors}]{\text{Positional embeddings}}$   $P(0), P(1), P(2), \dots \xrightarrow[\text{Model uses}]{\text{Model uses}}$   $E(\text{"the"}) + P(0), E(\text{"cat"}) + P(1), \dots$

**Two Main Types:**  
 Fixed (sinusoidal): based on sin/cos functions of position (as in the Attention is all you need paper)

Learned: Trainable vectors 1-2 - trained the same way as word (token) embeddings during training (used in GPT)

→ this way model knows which token came where