



PROMPT ENGINEERING FOR THE PREVENTIVE HEALTH COPILOT

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Objective:

Build an AI Copilot that gives personalized preventive health recommendations using multi-step reasoning and function/tool calling



Key Capabilities:

- Handles multi-turn queries (e.g., diet tips, reminders)
- Uses function-calling for tools like `get_diet_tip()` and `schedule_reminder()`
- Optimized through prompt iteration and evaluated using structured metrics



PROMPT STRATEGY AND EVOLUTION

Advanced Versions Enable Reasoning & Tools

Comparing prompt engineering strategies and capabilities across four versions

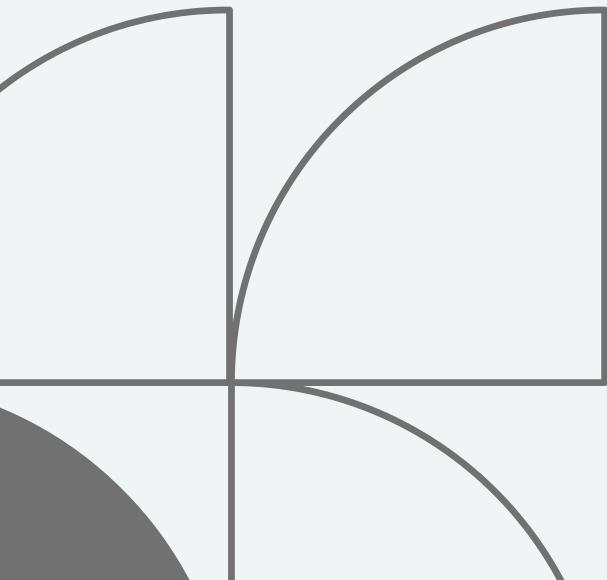
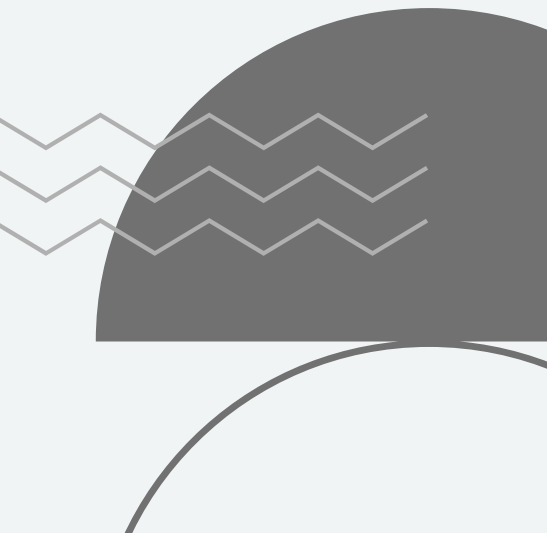
| Core Features | | | Capabilities | |
|---------------|--------------------|----------------------|--------------|------------|
| Version | Prompt Type | Strategy | Reasoning | Tool Usage |
| V1 | Simple Instruction | Direct Ask | ✓ | None |
| V2 | Few-shot Examples | Patterned Response | ✓ | None |
| V3 | ReAct | Chain-of-Thought | ✓ | Tools |
| V4 | Plan-Solve | Planning + Execution | ✓ | Tools |

1. Reasoning = Logical thinking in response
2. Tool usage = External tools integration



FUNCTION CALLING INTEGRATION



- various functions to enhance user interaction, including
`get_diet_tip(condition: str) → str`
`schedule_reminder(task: str, date: str) → str`.
 - Includes additional tools like `health_risk_assessment`,
`nutritionix_food API`, `openfoodfacts_nutrition API`,
`get_exercise_plan`.
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Evaluation Strategy

5 Pillars of Evaluation

Metrics to assess strategy effectiveness and quality

Evaluation focus

5 Metrics

Key aspects of strategy evaluation



Accuracy

Correctness of advice



Coherence

Logical flow of reasoning



Reasoning Used

Did it reason through steps?



Completeness

Did it fully answer the query?



User-Friendliness

Clarity and readability

1. Accuracy = Correctness of advice
2. Coherence = Logical reasoning flow

Scoring: 0–5 based on metric aggregation

Evaluation Tool: `score_response()` with keyword and tool usage validation

Results & Insights

ReAct Model Excels in Accuracy

Comparing four model versions by score, coherence, and reasoning

Performance Metrics

| Version | Score (0-5) | Accuracy | Coherence | Reasoning Used |
|-----------------|-------------|----------|-----------|----------------|
| V1 - Basic | 3.0 | ✓ | ✓ | ✓ |
| V2 - Few-Shot | 2.0 | ✓ | ✓ | ✓ |
| V3 - ReAct | 4.0 | ✓ | ✓ | ✓ |
| V4 - Plan-Solve | 3.0 | ✓ | ✓ | ✓ |

1. Accuracy = Correct outputs
2. Reasoning Used = Logical steps applied

Performance Evolves with Reasoning Tools

Comparing versions by scores and notes to track improvements

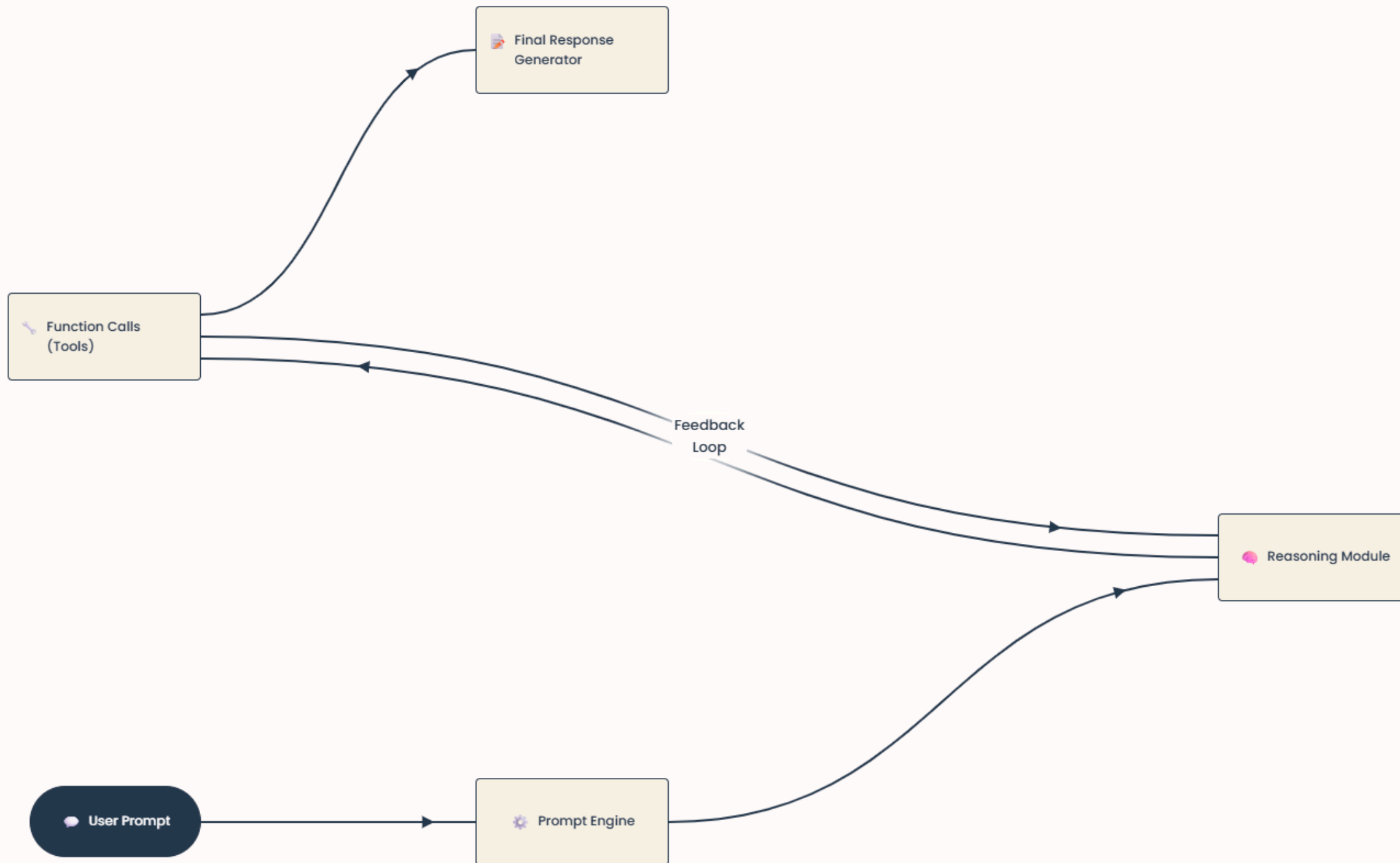
| Version | Score (/5) | Notes |
|---------|------------|--|
| V1 | 3 | Too basic, no dynamic reasoning |
| V2 | 2 | Static examples help a little |
| V3 | 4 | ReAct improves relevance |
| V4 | 3 | Best performance, tools + reasoning make a |

1. Score = Rating out of 5
2. Notes = Key observations per version

Insight:
Prompt engineering with Plan-Solve + Tool-Use yields the most reliable and intelligent responses

ARCHITECTURE & COMPONENTS

Illustrates key modules and their interactions in AI systems



- Modular agents using Agno and LLama LLM
- Tools exposed via function-calling
- Evaluated responses stored for comparison

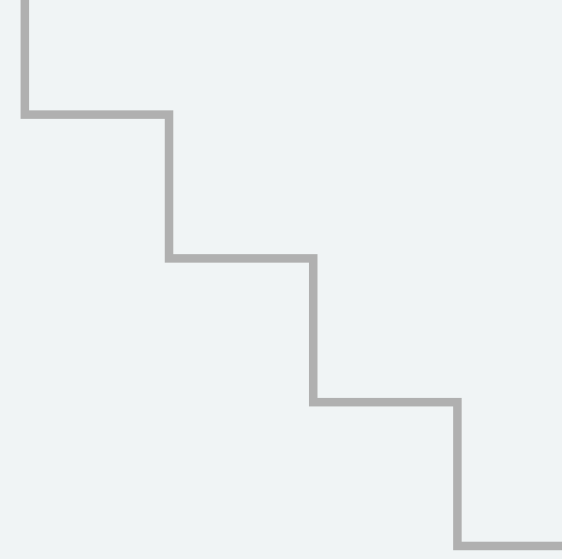
Deliverables & Repo

Deliverables:

- ☒ Prompt evaluation notebook with CSV results
- ☒ Agent + tools integrated in clean Python modules
- ☒ Streamlit UI (optional)
- ☒ Slide deck (this)

GitHub Repo Checklist:

- <https://github.com/Agaramsaikrishna/Health-Copilot.git>



**THANK
YOU**

