

SpindleAI Application

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MVP - 3 Agents + Virtual Tool Manager

3 Agents

- **Solver Agent** - Uses math tools and LLM to solve
- **CAS Agent** - Uses LLM to frame the problem to be solved by a Computer Algebra System
- **Verification Agent** - Uses LLM to solve

Tools

- **MathToolbox (Python)** sum, product, divide, subtract, power, sqrt, modulo, round_number
- **MathToolbox (Rust)** calculate_average
- **Virtual Tool Manager** - Manage functions

Model

- gpt-4o-mini
- temperature = 0

Agent Framework

- Langchain

Languages

- Python + Streamlit
- Rust

MVP - Keep it Simple

- Start with a simple 3 agents
 - Solver Agent
 - Uses MathToolbox's set of math functions
 - Uses LLM to guide an answer using MathToolbox
 - Verification Agent
 - Uses LLM to guide an thorough answer
 - CAS Solution Agent
 - Uses LLM to rephrase into a question consumable by a CAS
 - Uses a CAS to form an answer

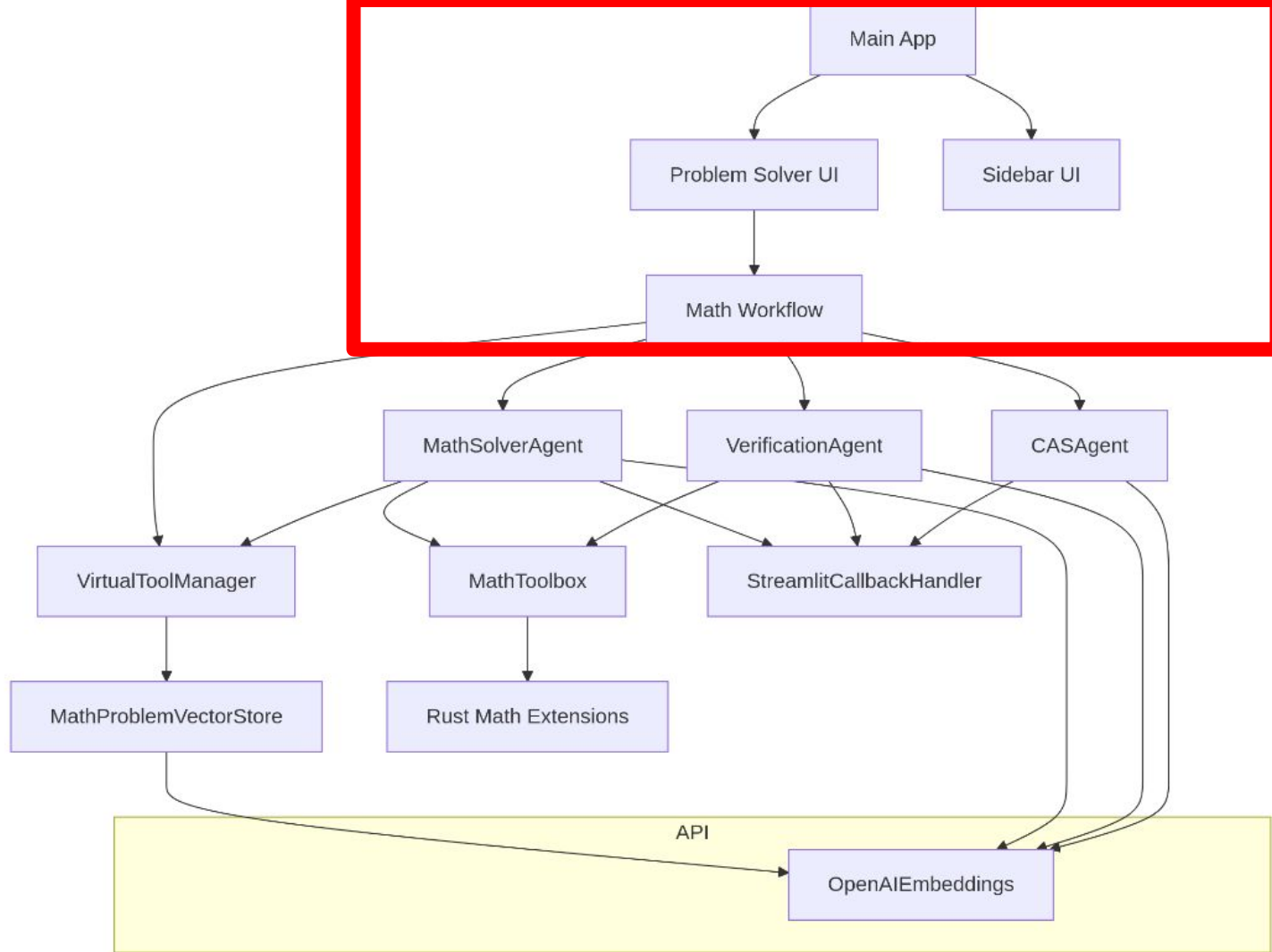


MVP - Majority Rules

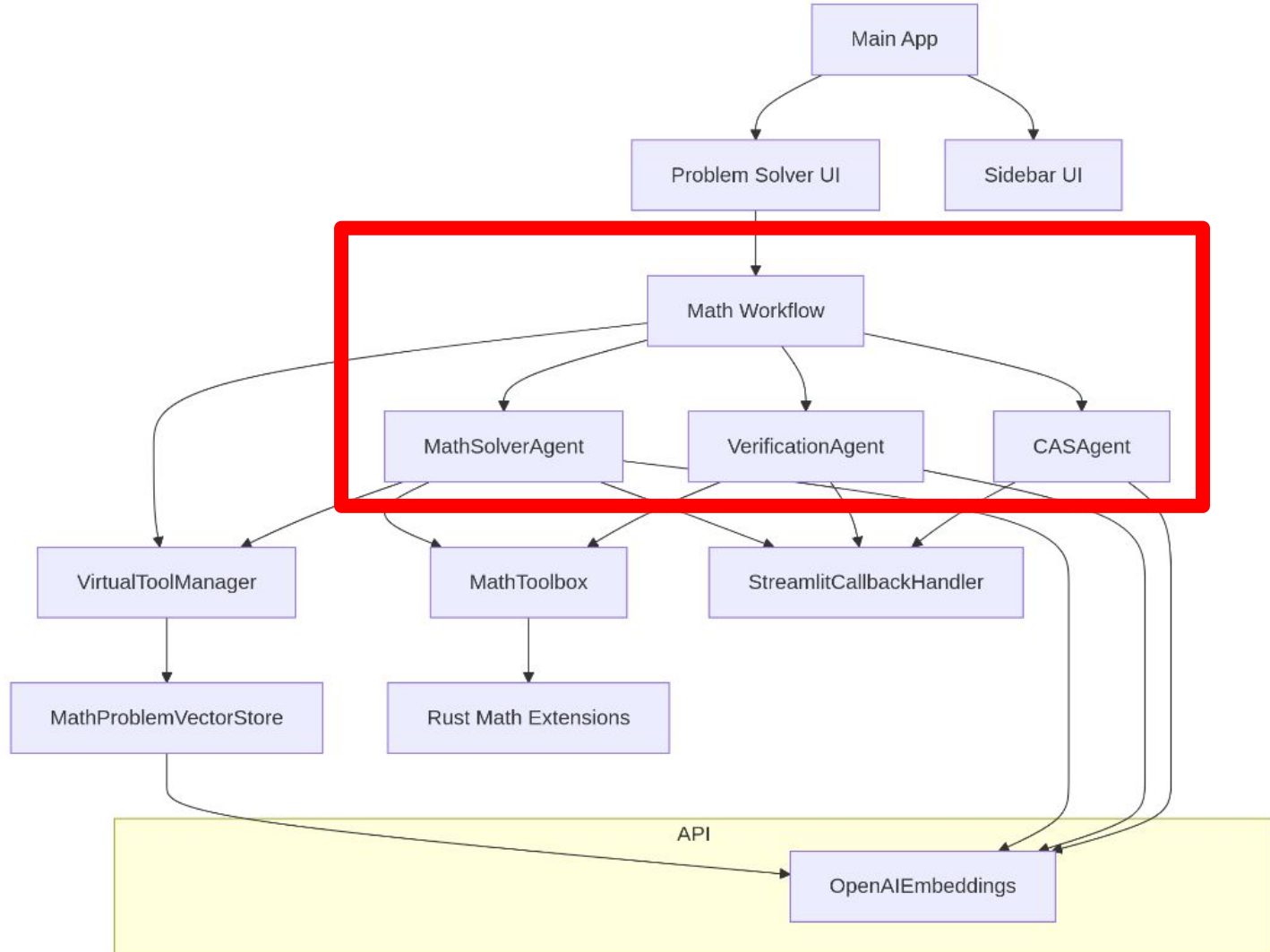
- 1. If any two or three agents agree on a solution, that solution is chosen as the majority solution.
- 2. If there's no majority agreement, we check two specific cases:
 - If the solver and CAS agents agree (but validation disagrees), we use their solution.
 - If the CAS and validation agents agree (but solver disagrees), we use their solution.
- 3. Only if we don't have any of the above cases, we fall back to the original behavior of using the solver solution with verification.



UI Drives a Math Workflow

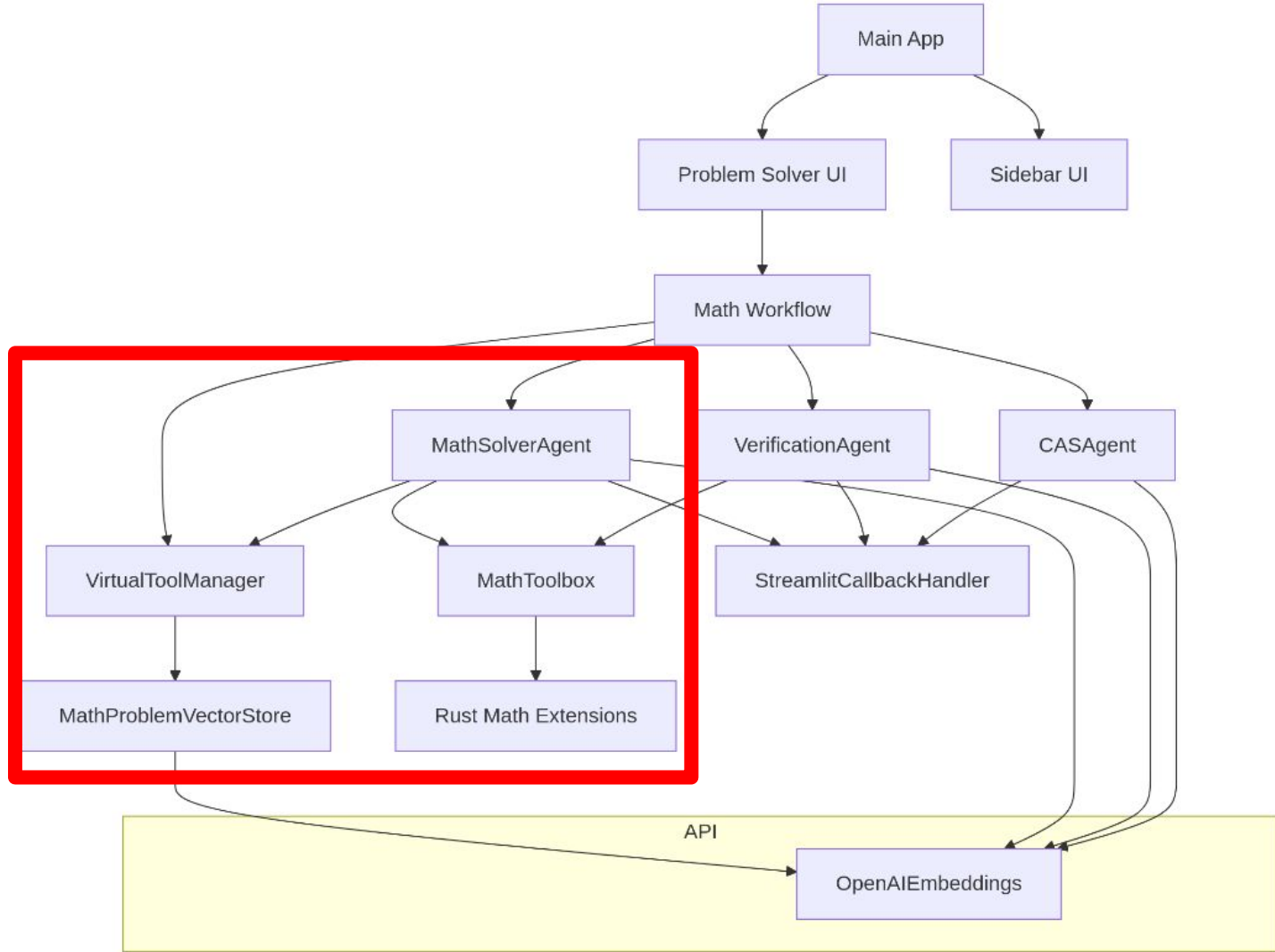


Math Workflow Orchestrates 3 Agents



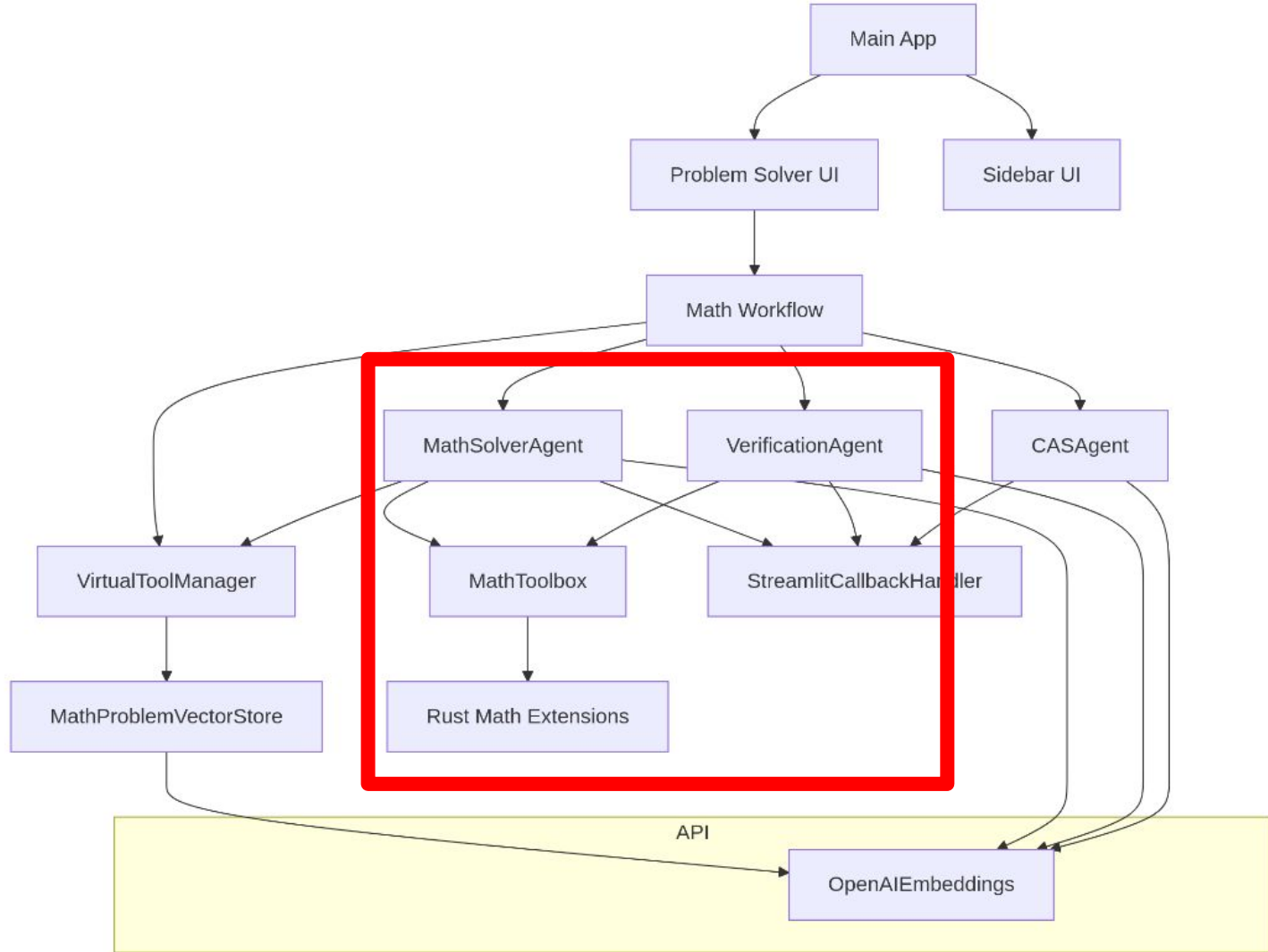
MathSolver Agents

uses Virtual Tool Manager to get relevant function



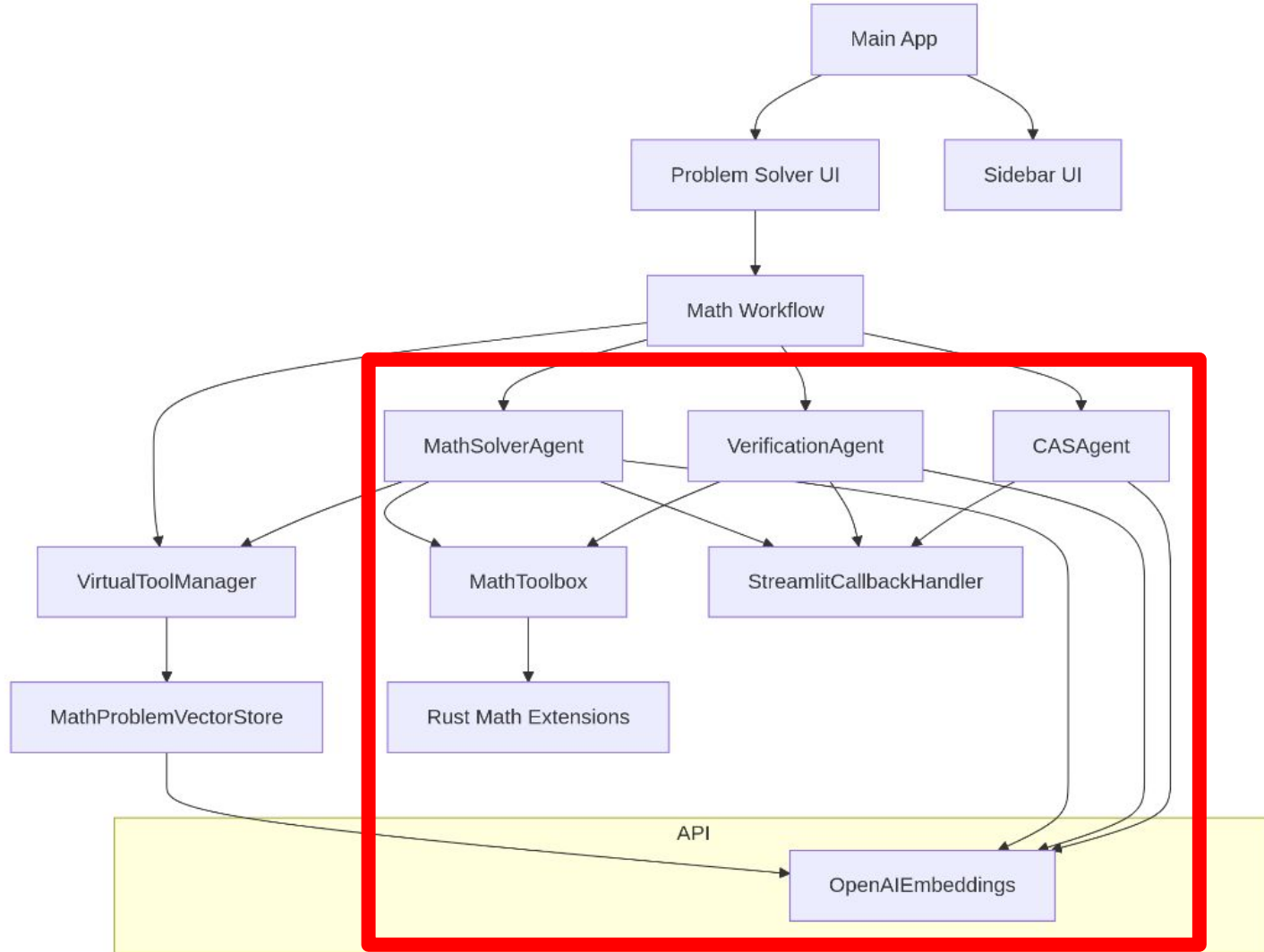
2 Agents

Rely on MathToolbox For Math Functions

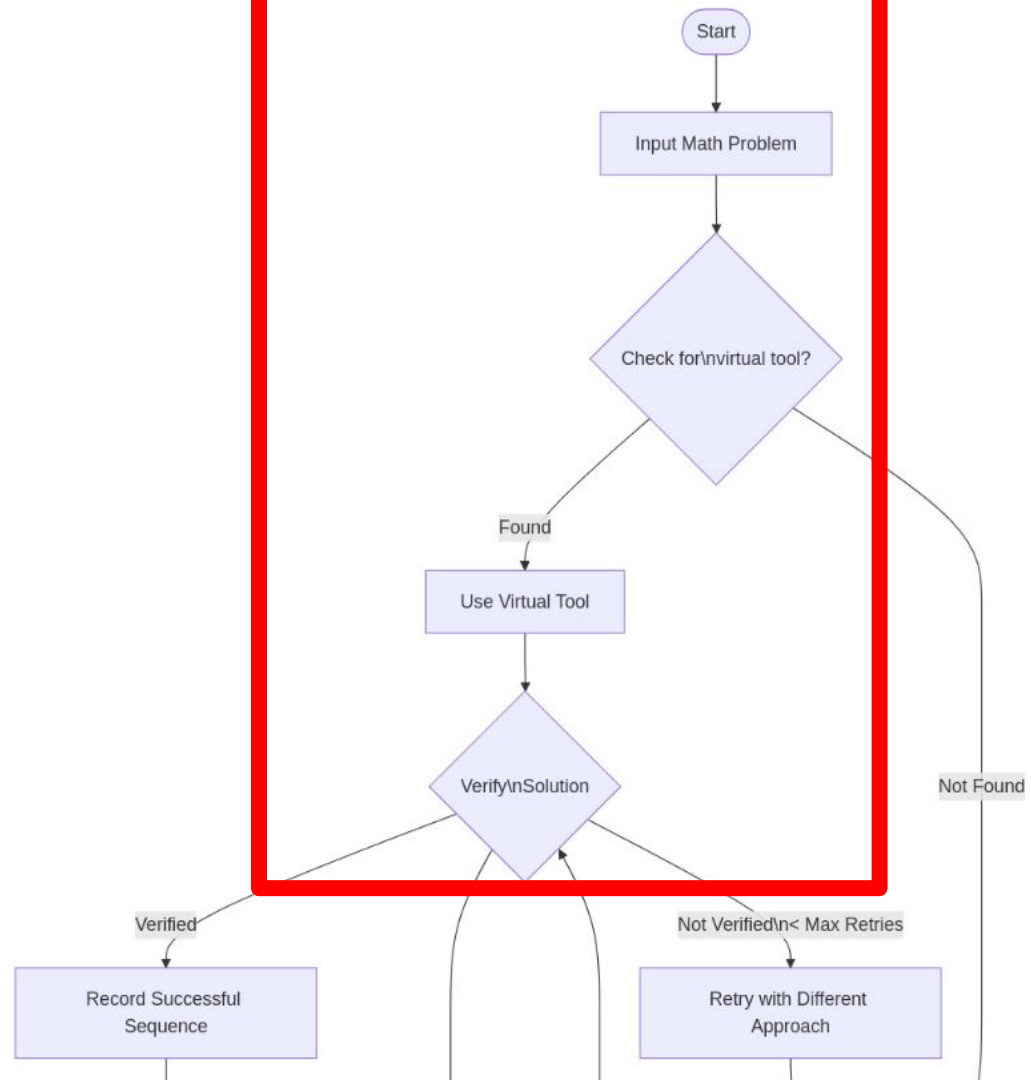


3 Agents

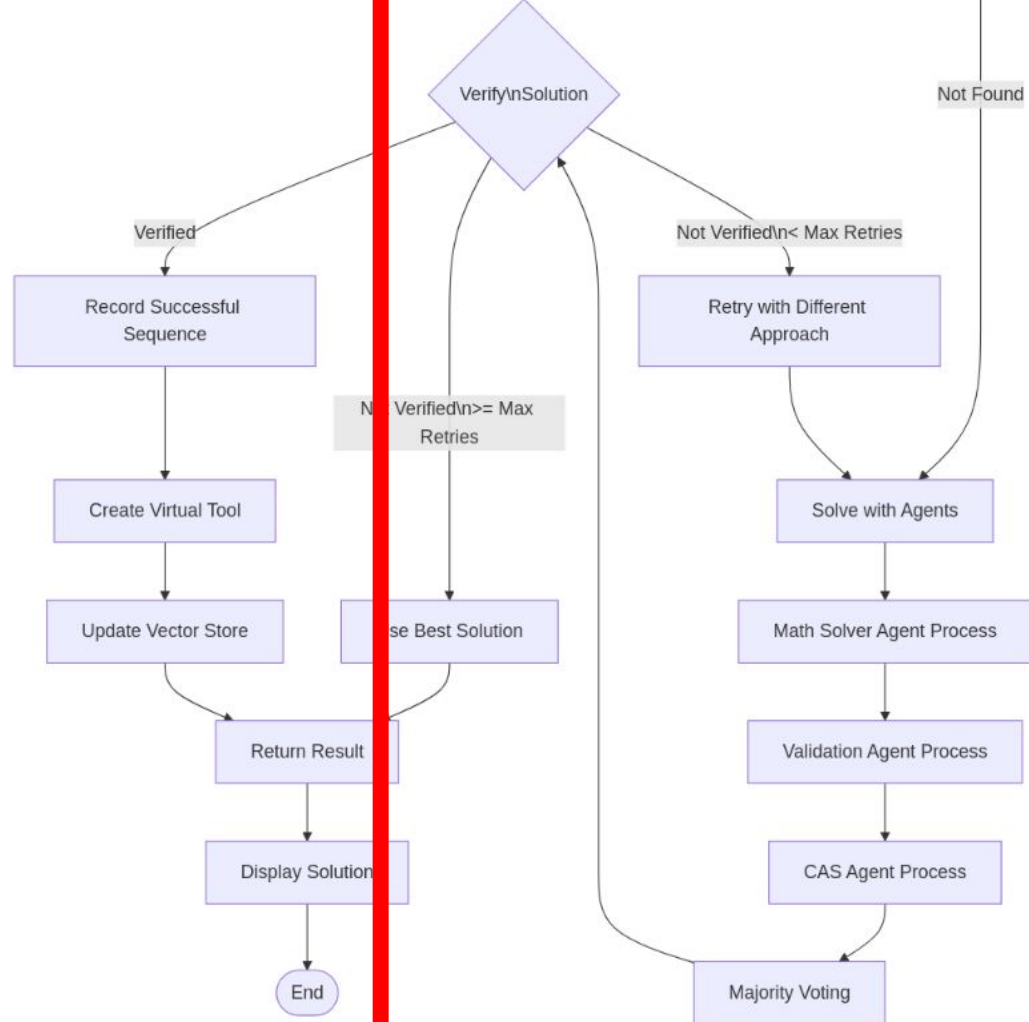
Rely on LLM
integration
to work with
the math
question



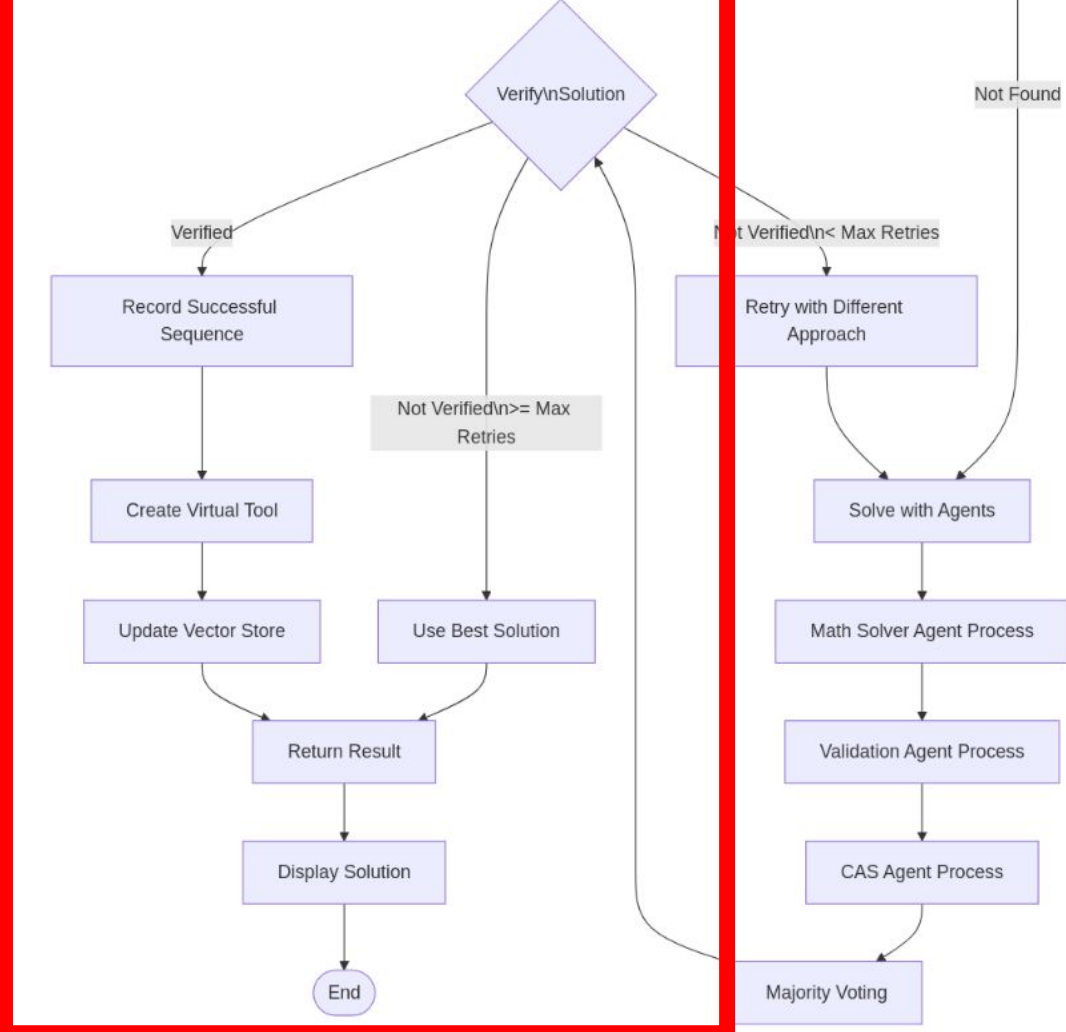
- Start by checking if there's a virtual tool.
- Use it if there is one
- Verify the Solution



- If verification fails
- Solve with Agents
- Use Majority Voting to verify answer



- Record successful Sequence
- Create a new virtual tool
- Update Vector Store
- Return/Display result



Demo - Python

Demo the python app using streamlit

- Live: <https://mvpmathsolver.replit.app/>
- Video: <https://youtu.be/Z3t02ZOIFfs>

The screenshot displays the MVP Math Solver web application interface, which is divided into two main panels: a left sidebar for settings and statistics, and a main content area for solving problems.

Left Sidebar:

- Settings:** Includes a toggle for "Add Tool Errors" (currently off).
- Tool Statistics:** A table showing the number of calls and error rates for various mathematical operations.

Tool	Calls	Error Rate
sum	0 calls	0.0%
product	2 calls	0.0%
divide	0 calls	0.0%
subtract	2 calls	0.0%
power	2 calls	0.0%
sqrt	0 calls	0.0%
modulo	0 calls	0.0%
round_number	0 calls	0.0%
- Virtual Tools:** A section for virtual tools, currently showing "VirtualTool_General_14dc969b: Solves General problems similar to: '(2-3)*5^2'". It also includes a sequence: `subtract → power → product`.

Main Content Area:

- Enter a Math Problem:** A section with a "Run Evaluation" button and a "Run Evaluation on Sample Problems" button.
- Sample Problems:** A section with a dropdown menu for selecting a sample problem (currently showing "(2-3)*5^2") and a "Solve Problem" button.
- Solution:** Displays the result of the evaluation, which is "-25".
- Verification:** A green box indicating that the solution has been verified as correct (Attempts: 1).

At the bottom of the main content area, a status message reads: "VERIFIED: The proposed solution gives the correct".

Github

Invited <https://github.com/orgs/spindle-app/people>

https://github.com/mtshomskyieee/mvp_math_solver

- Python
- Rust (ported from Python)

Thanks

- Next Steps
 - Scaleable API using FastAPI + Kubernetes
 - Port the demo to Rust

Appendix: Requirements

<https://spindle.notion.site/Coding-Project-Option-A-for-15757291437d804b87edf816a4212cdb>

This mission represents a stripped-down but realistic “toy version” of the kind of multi-agent system Spindle AI is engineering (including some actual challenges we’ve already faced):

- The Setup:** First, create ≥5 distinct, simple, deterministic tools that an LLM-based agent could call to help solve user-provided math problems (e.g. `SUM`, `DELTA`, `PRODUCT`, `QUOTIENT`, `MODULO`, `POWER`, `ABS`, `LOG`, `FIBO`, `SORT`, `AVG`, `MODE`, `ROUND`, `MIN`, `INTERPOLATE`, `DIFFERENTIATE`, `INTEGRATE`, `FACTORIAL`, ... — the specific tools are entirely up to you).
 - Modify 1-2 of the most basic tools to intentionally but silently throw errors (and/or silently give incorrect answers) 30%-50% of the time the tool is called** . You may also want to include a basic `GET_USER_INPUT` tool for requesting input/clarification from a human user. (You can organize all tools in some form of “toolbox” if you want, but we’d prefer you do **not** hardcode a string listing all the tools, their docs, and their usage examples in a *single* prompt file or prompt mega-string anywhere in the project.)
 - The Architecture:** Prototype a multi-agent system with **at least 2 agents** and *at most 5 agents* (for whatever definition of “agent” you believe makes sense in this context) that discovers which tools are available and sequences tool calls to **reliably** solve basic user-provided math problems (or if you prefer, mathy word problems). The agents can *only*****use the available tools (**including the unreliable tool[s]**) , i.e. no LLM-hallucinated arithmetic should be used for user-facing answers (even if that arithmetic is correct, as is increasingly the case among frontier models).
 - You might well choose to include a lightweight planning, reasoning, and/or task decomposition layer in your prototype — but unless you have a compelling justification, all *user-facing* outputs (and most intermediate outputs) should be structured or semistructured, not unstructured.
 - Don’t hesitate to ask us for an OpenAI API key or Anthropic API key.** Otherwise, we’re happy to reimburse these costs after submission (*within reason/at Spindle’s discretion*).
 - The Twist:** When your prototype identifies a sequence of tool calls that reliably or *fairly reliably* solves a certain class of math problem(s) **based on successful execution(s)** , it should learn to do something like (e.g.) **memoize or semantically cache that sequence of tool calls as a single, idempotent new `VirtualTool`** (i.e. some learning behavior akin to “bundling” the tool calls into a *single* new idempotent tool, to which a *single* call can be made, which can be reliably invoked *next time a math problem of the same or similar form is encountered*).
 - The Finish Line:** Prove programmatically that your prototype works reasonably well (or at least that it could be *completed* to work reasonably well, if short on time).
 - Bonus points for using actual evals to show this.**
 - (If you’re an “evals-focused” candidate, consider reframing/approaching the entire task through the lens of an evals system instead, i.e. evals-driven development. Just tell us to judge your quality vs. emphasis vs. completion accordingly.)**
 - Bonus Points:**
 - Create the math toolbox/interfaces in a non-Python language (ideally Rust, Go, or Typescript).
 - If you decide to use a vector database anywhere, consider prototyping your *own* vector DB or VDB-like utility. (Not if this takes up all your time, though. It’s not the most important part.)
- If you don’t have enough time for a project like this, or have alternate ideas, please let us know so we can find a path forward that we all feel good about** Either way, we really look forward to seeing you through these next steps.