

Documentation is Code (with agentic AI)

Allow billions of people to build
billions of solutions



You won't enjoy it on as many levels



What is DisC

- Documentation is Code(DisC)
- Any solution where instructions written using natural language are *executed* by an Agent

Working with Agentic AI

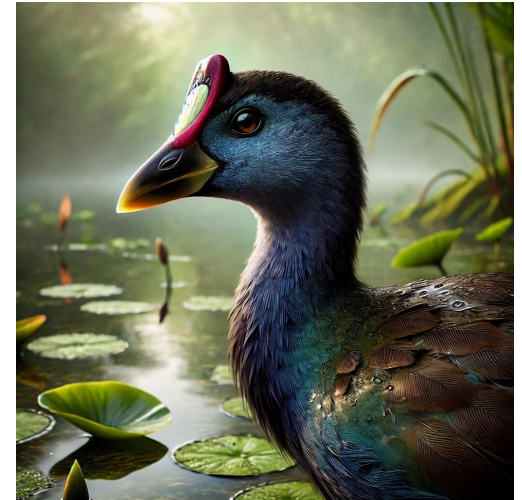
- What is an agent?
- High learning curve
- LLMs produce excessive output
- Inability to modify output
- No empathy, no mercy

Food for thought

- What if we are not looking for an answer?
- What if it is ok to make mistakes?
- Maybe Agents are not supposed to solve every problem?

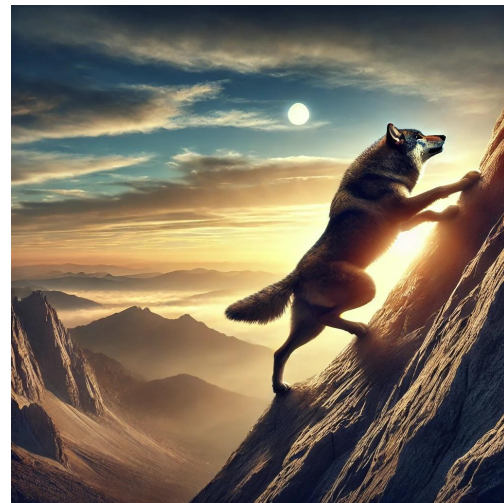
Evolution of Code

- Tubes, punch cards, assembly tapes
- Higher level languages to meta-frameworks
- No code
 - a GUI that builds things for you
 - Millions of products
- Documentation is Code
 - Just write the requirements and execute



Billions of people to build billions of solutions

- What about runtime performance?
- $f(\text{problem}) = \text{solution}$
 - The solution is a function of the problem
- Benefits outweigh the costs
 - Proof of Concept
 - Personal projects
 - Internal tooling
- Multi-modal zero code solutions



Anatomy of DisC

- End user will only use natural language
- Minimize software engineering concepts
 - Unstructured data
 - Implicit algorithms
- Dynamic tools - no code changes for adding tools
- Human in the loop built-in
- Consistent output
- Iterative development



Demo

- Build a dashboard for our github repo
 - The requirements keep changing
- Start with a traditional solution
- Then transform to a DisC solution



Requirements

- For the most prolific author of github issues, give me the number of issues they have created.
- Given the author's issues, I want to know the number of comments that same author has made.



Frontend

...

```
<div class="container mt-5">
  <table class="table table-striped">
    <tbody id="dataTable">
      <!-- Data will be populated here -->
    </tbody>
  </table>
</div>

<script>
  async function loadData() {
    const response = await fetch('http://localhost:5001/data');
```

...

comment_count	issue_count	name
3	2	yasonk

Backend

```
for issue in issues:
    author = issue["user"]["login"]
    author_count[author] = author_count.get(author,0) + 1

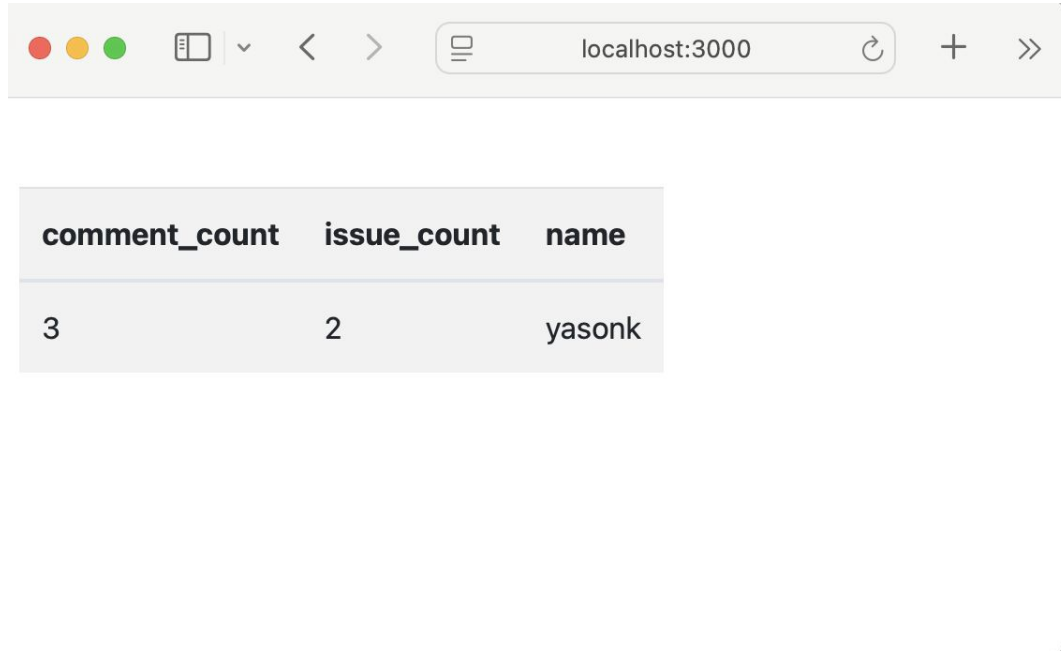
author = max(author_count, key=author_count.get, default=None)

issue_count = author_count.get(author, 0)

comments_count = sum(1 for comment in comments if comment["user"]["login"] == author)

result = {
    "data": [
        {"name": author, "issue_count": issue_count, "comment_count": comment_count}
    ]
}
return jsonify(result)
```

Result

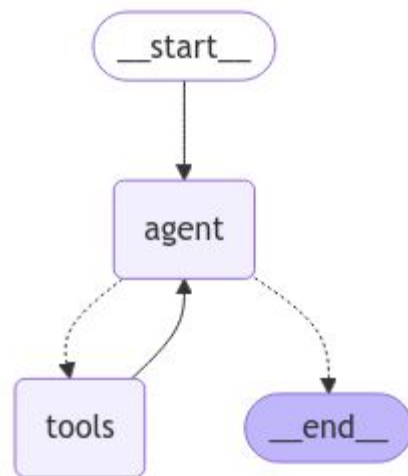


A screenshot of a web browser window. The address bar shows 'localhost:3000'. The main content area displays a table with three columns: 'comment_count', 'issue_count', and 'name'. The table has one data row with the values 3, 2, and yasonk respectively.

comment_count	issue_count	name
3	2	yasonk

ReAct Agent

- Reason - Act - Observe - Iterate loop
- ReAct : Synergizing Reasoning And Acting In Language Models <https://arxiv.org/pdf/2210.03629>
- Tool calling support
- Prebuilt LangGraph ReAct agent



ReAct Agent Implementation

```
# Precoded tools (not dynamic)
toolkit = GitHubToolkit.from_github_api_wrapper(GitHubAPIWrapper())
tools = [setattr(tool, "name", tool.mode) or tool for tool in toolkit.get_tools()]

llm = ChatOpenAI(model="gpt-4o", temperature=0)

def ai(query: str, response_format: Type[BaseModel]):
    agent_executor = create_react_agent(llm, tools, response_format=response_format)
    events = agent_executor.stream({"messages": [{"user", query}]}, stream_mode="values")

    event = {}
    for event in events:
        event["messages"][-1].pretty_print()

    structured_response = event.get("structured_response")
    return json.dumps({"data": [structured_response.model_dump()]})
```

Route Implementation

```
class Author(BaseModel):  
    name: str  
    issue_count: int  
    comment_count: int
```

```
@app.route('/data', methods=['GET'])
```

```
def get_data():
```

```
    query = """
```

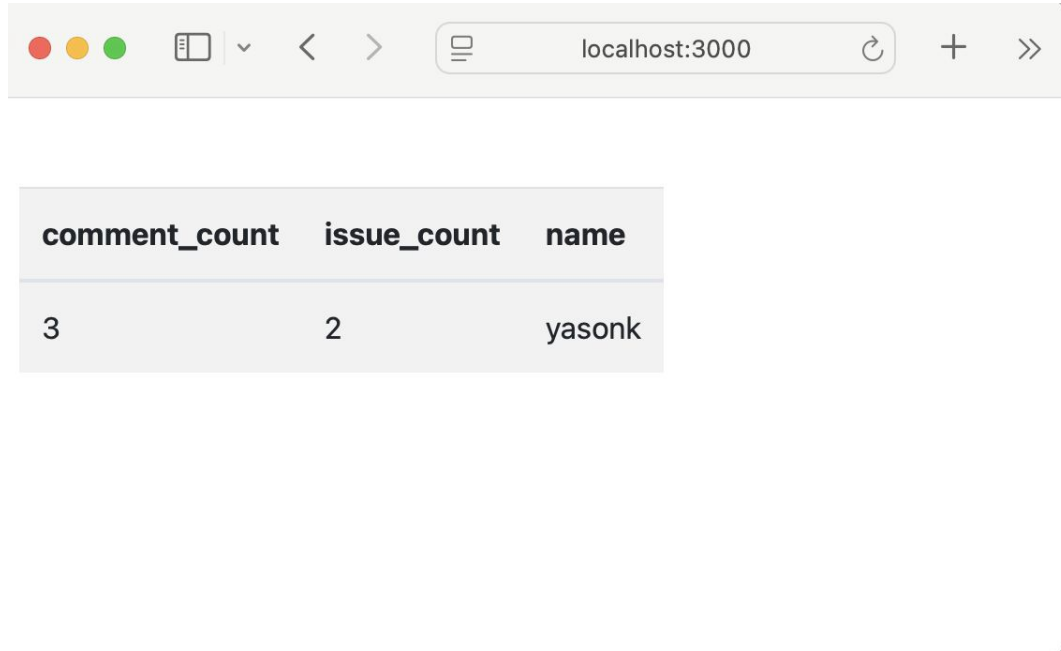
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For the most prolific author of github issues, give me the number of issues they have created.
```

```
Given the author's issues, I want to know the number of comments that same author has made.
```

```
""")
```

```
    return ai(query, response_format=Author)
```


Result



A screenshot of a web browser window. The address bar shows 'localhost:3000'. The main content area displays a table with three columns: 'comment_count', 'issue_count', and 'name'. The table has one data row with the values '3', '2', and 'yasonk' respectively.

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"""
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```

~~Dynamic Tools~~

~~Unstructured Data~~

~~Implicit Algorithms~~

~~Uses Natural Language Only~~

New Requirements

- Find the most prolific author of github issues, and I want to see all of their issues and comments
- I will need to see only the author, issue title, and comments.



Open Jupyter Notebook



Capabilities

- What did we achieve in 30 minutes?
 - End user will write only natural language
 - Unstructured data
 - Implicit algorithms



Limitations

- Tools still need to be coded
- Must learn agent specializations and limitations
- Runtime performance is slow
 - But should always be 100x faster to develop
- Inconsistent outputs
 - Blasts out overreaching solutions
 - Incremental refinement is challenging
- Engineering rigor
 - Security
 - Scalability
 - Debugging



Next Steps

- Tools to be dynamic without requiring new code
 - Simply drop in documentation for a tool
- Stateful agents
 - Checkpointing and persistence
 - Iterative development
 - Output consistency
- Improve human-in-the-loop processes
 - Human in the loop is unavoidable
 - Assist with debugging



Are we obsolete now?

- Expert knowledge is still needed
 - Expertise on building Agents
 - Communicating with AI agents
 - Retooling for AI (terminals, browsers)
 - Re-documenting for AI
- Need better LLMs



Conclusion

- With DisC, users will write their specification and the agent will run it
- Solve of problems rarely solved before
 - Developer setup instructions
 - A small website for your party
 - A tennis league at work
 - A behavior tracking app for you child
- Billions of people will be able to solve billions of highly personalized problems



**Link to this
presentation**



My Profile

