

Do Large Language Models Dream of Sockets?

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Context and goals



Lots of excitement on generative AI for

- Human languages, chat bots
- Image and video creation
- Programming assistance
- Search and documents

Cool, but not at the heart of things from a protocol or network engineer perspective

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What if LLMs were able to also converse natively in protocol messages?

- There's multi-modal generative AI and support for multiple languages
- Could we "speak" protocols, too?

Related Work



arXiv > cs > arXiv:2402.19155

Computer Science > Machine Learning

[Submitted on 29 Feb 2024]

Beyond Language Models: Byte Models are Digital World Simulators

Shangda Wu, Xu Tan, Zili Wang, Rui Wang, Xiaobing Li, Maosong Sun

Traditional deep learning often overlooks bytes, the basic units of the digital world, where all forms of information and operations are encoded and manipulated in binary format. Inspired by the success of next token prediction in natural language processing, we introduce bGPT, a model with next byte prediction to simulate the digital world. bGPT matches specialized models in performance across various modalities, including text, audio, and

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PROSPER: Extracting Protocol Specifications Using Large Language Models

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Abstract

We explore the application of Large Language Models (LLMs) (specifically GPT-3.5-turbo) to extract specifications and automating understanding of networking protocols from Internet Request for Comments (RFC) documents. LLMs have proven successful in specialized domains like medical and legal text understanding, and this work investigates their

1 Introduction

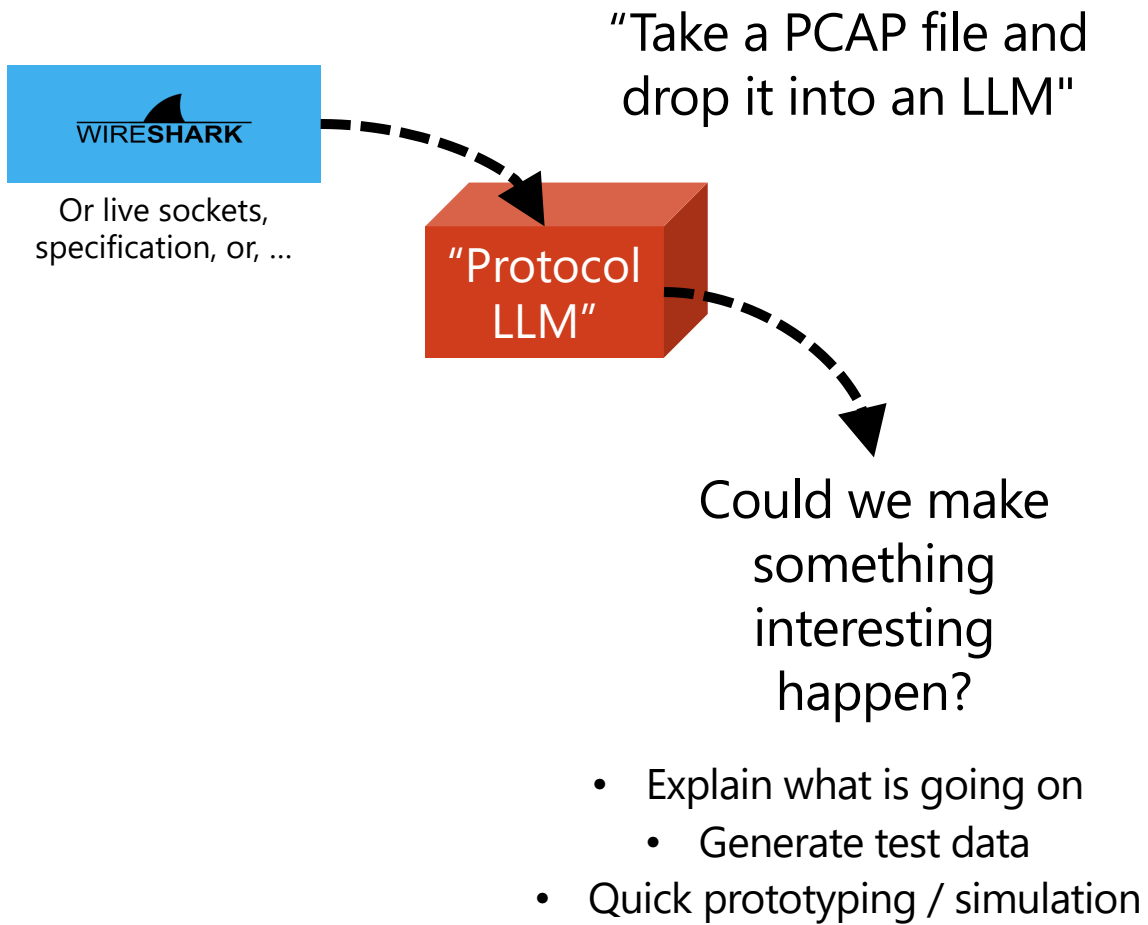
Network protocols serve as the foundation for communication between devices and systems but often are complex and diverse, making manual analysis and implementation time-consuming and error-prone. A common way of specifying network protocols is using request for comments (RFC) documents. Automatic protocol understanding from RFCs

Can you generate a byte sequence that represents a DNS message for to query the IPv4 address of arkko.eu?

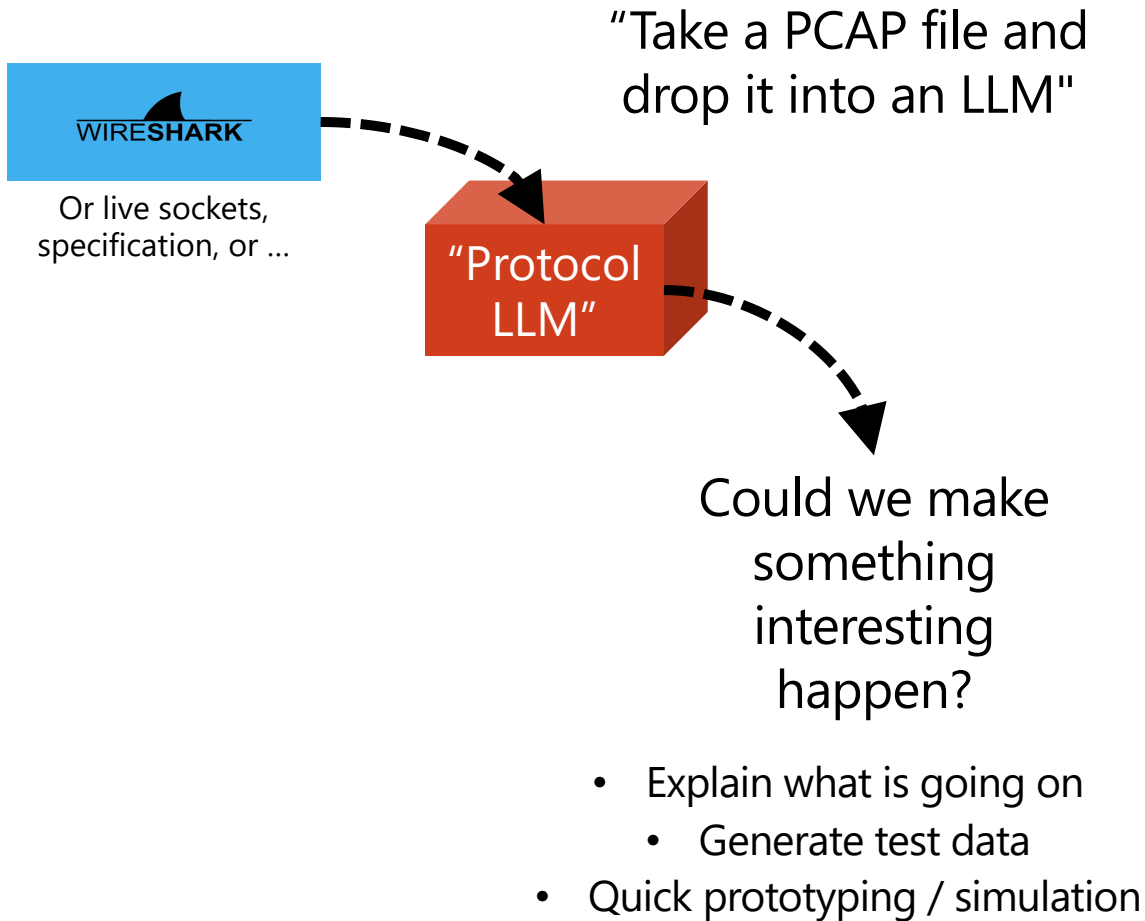
Here's a simplified example of what the byte sequence might look like for a DNS query for the A record of arkko.eu. This example assumes a transaction ID of 0x1234 and uses the QDCOUNT field to indicate one question.

1234 0100 0001 0000 0000 0000 0000 0561 726b 6b6f 0265 7500 0001 0001

Vision



Vision



Research Approach

Try to understand if this is feasible, and to what extent

- Test different use cases and techniques
- Experiment to quantify suitability in different use cases and the performance of different techniques

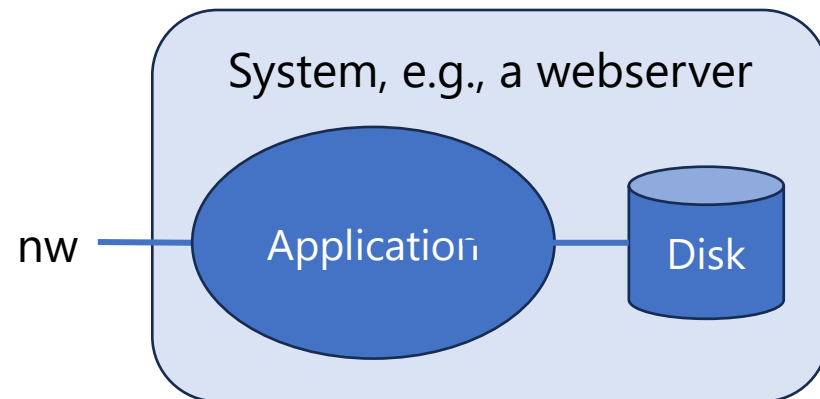
Early / in progress

Some Challenges



- **Complex fields** – length, checksum, encryption, ...)
- **Protocols are not everything** – real system behavior is not explained by protocols only
- **Security and safety** – reading logs or sending messages, accessing local resources
- **Hallucination** – correctness
- **Efficiency** – cost, energy, speed

As an AI, I'm unable to perform real-time calculations or generate dynamic content such as calculating a UDP checksum for a specific packet.



Some (Partial) Solutions

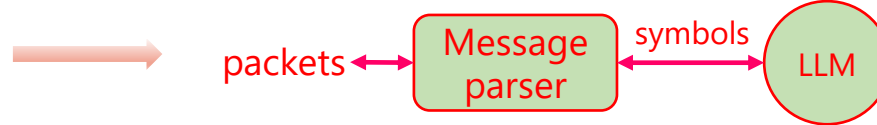


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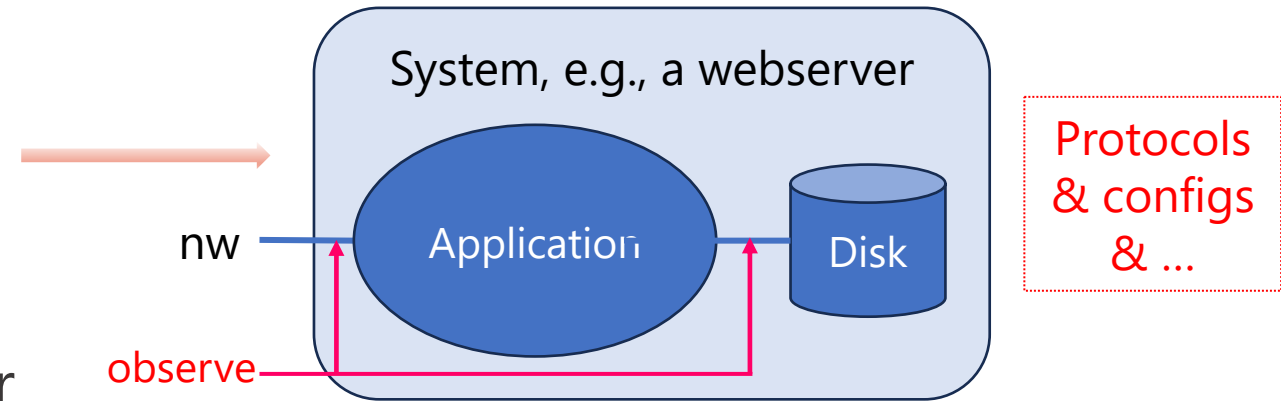


Combining traditional tools and AI

Some (Partial) Solutions



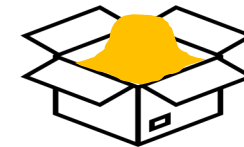
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Limits

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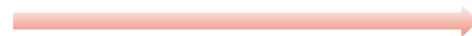


The right
use cases

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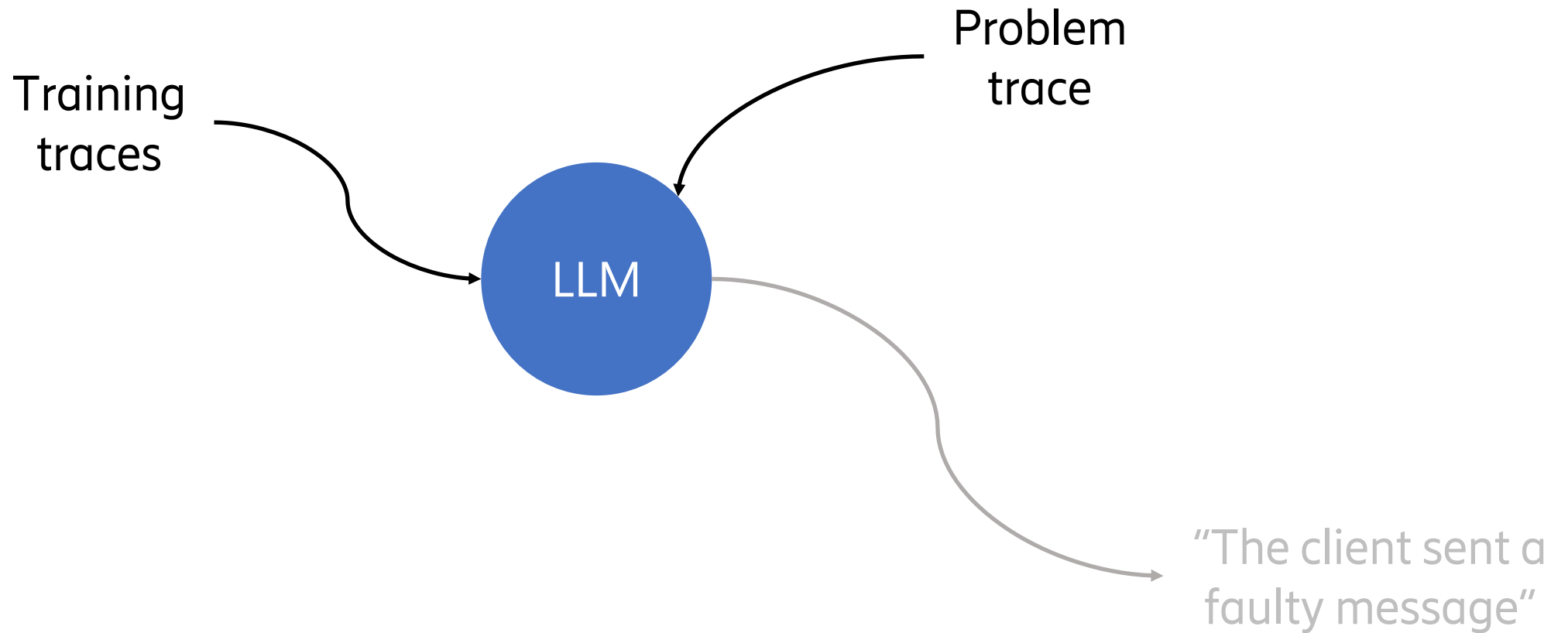
Code generation



Example Use Case: Diagnostics



Use Case Context: Training traces & Problem traces



strace.apache.5.pcap

Apply a display filter ... <=>

No.	Time	Source	Destination	Protocol	Length	Info
1	0.000000	172.17.0.1	172.17.0.2	TCP	80	59304 → 80 [SYN] Seq=0 Win=64240 Len=0 MSS=1460 SACK_PERM TSval=1122461665 TSecr=0 WS=128
2	0.000012	172.17.0.2	172.17.0.1	TCP	80	80 → 59304 [SYN, ACK] Seq=0 Ack=1 Win=65160 Len=0 MSS=1460 SACK_PERM TSval=2963594550 TSecr=1122461665
3	0.000041	172.17.0.1	172.17.0.2	TCP	72	59304 → 80 [ACK] Seq=1 Ack=1 Win=64256 Len=0 TSval=1122461665 TSecr=2963594550
4	7.857502	172.17.0.1	172.17.0.2	HTTP	89	Continuation
5	7.857547	172.17.0.2	172.17.0.1	TCP	72	80 → 59304 [ACK] Seq=1 Ack=18 Win=65152 Len=0 TSval=2963602408 TSecr=1122469523
6	7.857839	172.17.0.2	172.17.0.1	HTTP	478	HTTP/1.1 400 Bad Request (text/html)
7	7.857879	172.17.0.1	172.17.0.2	TCP	72	59304 → 80 [ACK] Seq=18 Ack=407 Win=64128 Len=0 TSval=1122469523 TSecr=2963602408

> Frame 4: 89 bytes on wire (712 bits), 89 bytes captured (712 bits)
 > Linux cooked capture v2
 > Internet Protocol Version 4, Src: 172.17.0.1, Dst: 172.17.0.2
 > Transmission Control Protocol, Src Port: 59304, Dst Port: 80, Seq: 1, Ack: 1, Len: 89
 > Hypertext Transfer Protocol

File Data: 17 bytes
 Data (17 bytes)
 Data: 484145202f696e6465782e68746d6c0d0a
 [Length: 17]

```

0000 08 00 00 00 00 00 00 16 00 01 00 06 02 42 24 27 .....B$'
0010 d5 bb 00 00 45 00 00 45 e5 e1 40 00 40 06 fc ab .....E..E..@.@...
0020 ac 11 00 01 ac 11 00 02 e7 a8 00 50 cc 25 e4 e6 .....P.%..
0030 98 a3 78 19 80 18 01 f6 58 5d 00 00 01 01 08 0a ...x.....X].....
0040 42 e7 86 93 b0 a4 dd 36 48 41 45 20 2f 69 6e 64 B.....6 HAE /ind
0050 65 78 2e 68 74 6d 6c 0d 0a                      ex.html..
  
```

AI:

“Due to the unrecognized or invalid HTTP method (“HAE”), the server responds with a “400 Bad Request” status code.”

Understanding diagnostics performance

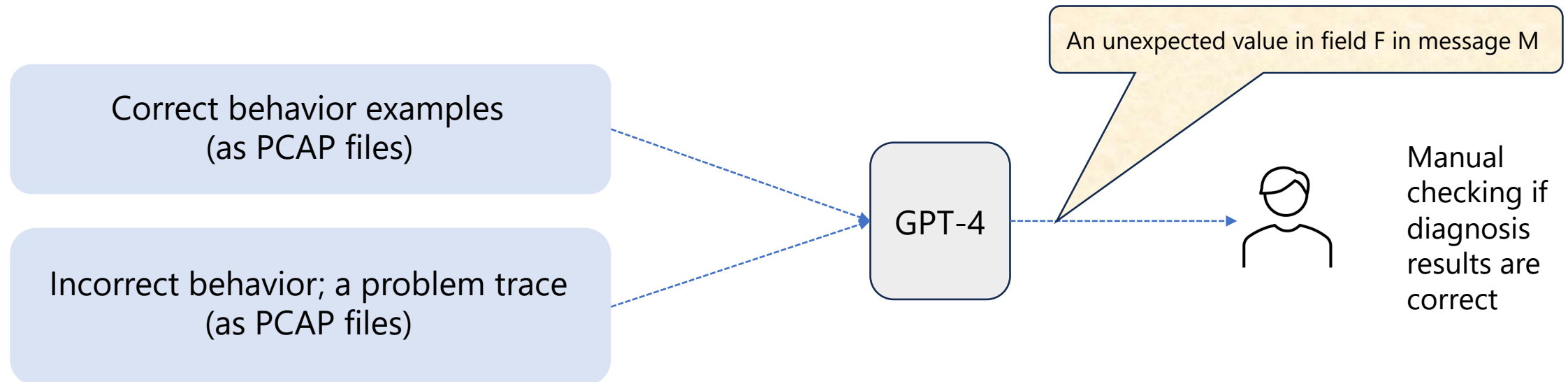


Could we quantify how good LLMs are in this?

We created a set of 78 different messages for a simple, artificial example protocol

We test the ability of the LLM to correctly identify if something was wrong with the messages

- Human determines if the LLM's explanation was reasonable



Test Results



Measure	Diagnosis results	
	Worst approaches	Best approaches
Issues correctly detected	70-80%	90-100%

Results vary depending on techniques used, protocol in question, tests, interpretation, and even runs

Conclusion: diagnosis seem feasible

Good results with either:

1. Input = training & problem traces (in parsed form)
2. Input = specification & problem trace

More work needed – these are only initial tests

Other Results



Simulate/replicate systems

We recorded Apache's behavior on HTTP and file system call interfaces

The LLM learned to itself behave like a server and it responded to messages on sockets, read files, ...

E.g., that a "GET /foo.html" message should lead to opening file "/var/www/foo.html"

Including when to generate 404s, how the number of read bytes should influence Content-Length value, etc.

Difficult to use as a real service due to reliability (hallucination), but perhaps useful for simulation/quick prototyping



Conclusions



We've found this exciting

Protocol and system behavior patterns is a good topic for LLMs

Feasibility for different use cases to be determined

It is important to apply LLMs for the right tasks, not necessarily every task

Plenty of research problems to look into, e.g., better understanding of diagnostics performance, complex protocols, different training methods, security, etc.

