# Can you summarize large docs such as PDF files on the web? Or do I need to upload them here for you to analyze them?

#### 🚫 🛛 🕺 Thought for 8 seconds 🔨

Okay, the user is asking whether I can summarize large documents like PDFs that are on the web, or if they need to upload them here for analysis. Let me think about this step by step.

First, I know that currently, I can't access external websites or documents directly. So if the PDF is online, the user would need to download it and then upload it here for me to process. But wait, can I even process uploaded files? Let me recall the capabilities. From what I remember, the user can upload files, and I can read the text from them if they're in a supported format like PDF, DOCX, TXT, etc. But there's a size limit, right? I should mention that there's a limit on file size, maybe 50MB or something like that. Also, if the PDF is scanned or image-based, I can't read that unless it has OCR text. So I need to clarify that they need to upload the file here, and I'll analyze the text content. Also, maybe mention that for very long documents, the summary might be concise and they can ask for specific sections if needed. Let me structure this in a clear way. Start by confirming they need to upload the file, explain the supported formats, size  $\checkmark$ , and text extraction limitations. Then reassure that once uploaded, I can summarize or answer questions about the content. Also, suggest they can provide key excerpts if uploading isn't possible. Make sure the tone is helpful and clear.

Currently, I cannot directly access or analyze documents (like PDFs) that are hosted externally on the web. To summarize or analyze such files, you would need to **upload the document directly here** (via the paperclip/attachment icon in the chat interface).

# Key Notes:

- 1. Supported Formats: PDF, DOCX, TXT, and others (depending on platform).
- 2. Size Limits: Typically up to 50MB (varies by platform).
- 3. Text Extraction:
  - Works best for text-based PDFs.
  - Scanned/image-based PDFs require OCR (Optical Character Recognition) tools first.
- 4. Long Documents: I can summarize key points, but very long files may require focusing on specific sections.

If you can't upload the file, feel free to share **key excerpts or questions** from the document, and I'll help analyze those!

6209



I uploaded a 12 page PDF file. Please summarize its contents for me.

#### 🔇 🕺 Thought for 13 seconds 🔨

Okay, let's see. The user uploaded a 12-page PDF about using Google Gemini and Anthropic's Claude to analyze reverse-engineered code. My job is to summarize this.

First, I need to go through each page of the document. The main topic is how these AI tools handle disassembled and decompiled code. The author, Andrew Schulman, is testing their capabilities in reverse engineering contexts.

Page 1 introduces the experiment: using Gemini to analyze an assembly listing generated from C code. Gemini correctly identifies the code's purpose as regression analysis but makes a minor error about performance optimization. Page 2 shows that after removing identifiers, Gemini's accuracy drops, missing non-linear regression aspects.

Page 3 details Claude's analysis of the same code. Claude provides a more accurate breakdown, noting multiple regression types and R-squared calculations. The author compares Claude's introspection ability, which is more detailed than Gemini's.

Page 4 discusses testing with a Windows DLL disassembly. Gemini summarizes the DLL's functions and dependencies. When provided with a header file, Gemini explains how it aids analysis but gives generic examples until pushed for specifics.

Pages 5-7 continue with DLL analysis. Gemini uses the header to interpret assembly instructions, though initially hypothetical. The author notes Gemini's reliance on provided context without external searches.

Pages 8-9 cover decompiled Java from an Android app. Gemini identifies the code's purpose related to X.500 names and SpongyCastle, though obfuscated names require prompting to link to known functions.

Pages 10-12 show Claude analyzing a disassembled Windows executable. Claude infers the program tests multiple regression models, identifies best fits, and discusses limitations. The author compares Claude's deeper analytical approach.

The conclusion highlights both Als' strengths: Claude's detailed reasoning vs. Gemini's broader summaries. The author emphasizes Al's potential in reverse engineering despite current limitations.

I need to structure this summary clearly, highlighting key experiments, tools used, findings, and comparisons between Gemini and Claude. Make sure to note the different code types analyzed and the varying AI performance in each case.

Here's a concise summary of the 12-page PDF analyzing how **Google Gemini** and **Anthropic's Claude** examine reverse-engineered code:

# **Key Findings & Experiments**

#### 1. Assembly/Disassembled Code Analysis

- Google Gemini accurately inferred the purpose of a compiler-generated assembly file (regression analysis) but misinterpreted optimizations as intentional assembly coding. When identifiers (e.g., "regression") were removed, Gemini's accuracy dropped, incorrectly narrowing the scope to linear regression.
- Claude outperformed Gemini, identifying multiple regression types (linear, logarithmic, exponential, power) and explaining how it inferred functionality from variable/function names and data structures.

# 2. Windows DLL Disassembly

- Gemini analyzed txfw32.dll (a transaction-logging library), identifying dependencies and core functions (e.g., TxfLogCreateFileReadContext). With a header file, Gemini mapped assembly instructions to structures (e.g., TXF\_ID) but relied on hypothetical examples until pressed for specifics.
- Claude demonstrated deeper analytical skills, reverse-engineering a regression program's flow from disassembled code, including error handling, model comparisons, and R<sup>2</sup> calculations.

# 3. Decompiled Java Code

Gemini analyzed an obfuscated IETFUtils.java file (from SpongyCastle/Bouncy Castle),

identifying its role in X.500 name parsing and ASN.1 encoding. It linked obfuscated function names (e.g., n79411a) to known utilities (e.g., rDNsFromString) when prompted.

#### 4. Limitations & Observations

- AI Strengths: Both tools excelled at extracting high-level meaning from low-level code (e.g., inferring regression types from FPU operations, mapping DLL functions to system behaviors).
- Weaknesses:
  - Gemini occasionally defaulted to generic summaries without deeper analysis.
  - Neither AI proactively cross-referenced public documentation (e.g., searching for TxfLog APIs) unless explicitly guided.
- Reverse Engineering Insight: The AIs mirrored human reverse engineering by inferring higherlevel logic (e.g., "what process generated this data?") from low-level code patterns.

# Conclusion

- Google Gemini: Effective for quick summaries of disassembled/decompiled code but benefits from user guidance and supplemental context (e.g., headers).
- Anthropic Claude: More robust at reasoning through code logic, explaining inferences, and handling ambiguity (e.g., identifying multiple regression models).
- Future Potential: Al tools could revolutionize reverse engineering by automating tedious code analysis, though human oversight remains critical for accuracy.

For deeper analysis, focus on specific sections (e.g., Claude's regression breakdown, Gemini's DLL mapping) or ask follow-up questions!

9000



## Please tell me about the file I'm uploading as ietfutils\_java.pdf

#### ⊠ Thought for 15 seconds ∧

Okay, the user uploaded a PDF file named ietfutils\_java.pdf, which contains decompiled Java code from the SpongyCastle library. Let me start by understanding what this file is about.

First, I notice the package name: org.spongycastle.pl419a.pl437r.pl438a. SpongyCastle is known to be a repackaged version of BouncyCastle for Android, so this is likely related to cryptography. The class is named IETFUtils, which suggests it's handling something related to IETF standards, possibly X.509 certificates or ASN.1 encoding.

Looking at the imports, there are references to ASN10bjectIdentifier, RDN, X500NameStyle, etc. This points towards X.500 Distinguished Names, which are used in digital certificates. The methods here probably deal with parsing, formatting, or encoding these names according to IETF standards.

The code has methods like unescape, m79411a (renamed from a), which seems to parse a string into RDNs (Relative Distinguished Names). There's also a method m79406a that looks up ASN.1 Object Identifiers (OIDs) from a hashtable, which is common in handling certificate attributes.

There's a lot of string manipulation, especially dealing with escaping and unescaping characters, which is typical when working with Distinguished Names that have special characters. Methods like m79415e handle encoding of ASN.1 values, including hexadecimal representations.

The presence of methods like canonicalize and stripInternalSpaces indicates efforts to normalize strings according to specific formatting rules, which is crucial for consistent processing and comparison of certificate data.

I also see error handling for badly formatted directory strings, which suggests robustness in parsing inputs. The use of vectors and hashtables aligns with managing collections of OIDs and their string representations.

Considering the original analysis from the previous PDF, this class is part of utilities for processing X.500 names, aiding in tasks like parsing certificate subject names, validating OIDs, and ensuring proper encoding/decoding according to ASN.1 standards.

The obfuscated method names (like m79411a) are a result of decompilation, where original meaningful names are lost. However, based on functionality, these methods can be mapped to known BouncyCastle/SpongyCastle utilities, such as parsing RDNs from strings or handling OID lookups.

In summary, this IETFUtils class is a critical component for handling the intricacies of X.500 Distinguished Names and ASN.1 encoding within the SpongyCastle library, providing essential utilities for cryptographic operations involving digital certificates.

Here's a concise analysis of the **ietfutils\_java.pdf** file, which contains decompiled Java code from the **SpongyCastle** cryptography library (a repackaging of Bouncy Castle for Android):

# **Key Components & Functionality**

- 1. Class Purpose:
  - The IETFUtils class handles X.500 Distinguished Names (DNs) and ASN.1 encoding/decoding, crucial for parsing and formatting digital certificate data (e.g., SSL/TLS certificates).
- 2. Core Methods:
  - m79411a (renamed from a): Parses a string into Relative Distinguished Names (RDNs) by tokenizing input (e.g., "CN=Alice, 0=0rg") and constructing X.500 name components.
  - m79406a : Maps string identifiers (e.g., "OID.2.5.4.3") to ASN.1 Object Identifiers (OIDs) using a hashtable, critical for attribute validation.
  - $\circ$  unescape : Handles escaping/unescaping special characters (e.g.,  $\chi$  , " ) in DNs to comply with IETF formatting rules.
  - m79415e : Encodes ASN.1 values (e.g., strings, hex data) into human-readable formats, including handling DER encoding for non-string types.

# 3. String Manipulation:

- canonicalize and stripInternalSpaces : Normalize strings by removing extraneous spaces and ensuring consistent formatting for certificate comparisons.
- bytesToString : Converts hex-encoded byte arrays to strings for ASN.1 primitive decoding.
- 4. Error Handling:
  - Validates input formats (e.g., throws IllegalArgumentException for malformed directory strings like "CN=, 0=0rg" ).
- 5. Dependencies:
  - Relies on SpongyCastle classes like X500NameStyle , RDN , and ASN10bjectIdentifier for cryptographic operations.

# **Obfuscation & Reverse Engineering**

- Decompilation Artifacts:
  - Method names (e.g., m79405Y, m79411a) are obfuscated but map to original Bouncy Castle utilities (e.g., hex conversion, RDN parsing).
  - Code structure aligns with Bouncy Castle's IETFUtils class, which handles X.500 DN parsing/formatting.
- Key Identifiers:
  - ASN1String, DERUniversalString: Indicates ASN.1 data type handling.
  - X500NameBuilder , X500NameTokenizer : Tools for constructing and tokenizing X.500 names.

## Use Cases

- Digital Certificates: Parsing subject/issuer DNs in X.509 certificates.
- PKI Operations: Validating OIDs, encoding/decoding ASN.1 structures for cryptographic protocols.
- Data Normalization: Ensuring consistent string formatting for certificate comparisons.

## Limitations

- Obfuscation Challenges: Original method/variable names are lost, requiring inference from context.
- Dependency on External Libraries: Requires SpongyCastle/Bouncy Castle for full functionality.

This utility is foundational for cryptographic operations involving certificate processing, emphasizing compliance with IETF standards. Let me know if you need deeper analysis of specific methods!

6 C 🌔 🖓

What if I wanted an AI chatbot such as yourself to do the kind of analysis you just did for me on the ietfutils java file, but I need it to all happen on a local computer, not connected to the internet. What would you suggest for such as local/remote/airgrapped system? It would need to support back-and-force chats like we are having now, but all have in-depth ability to analyze code in at least C/C++, Java, and Python. It would need to support projects with not just a single file like we've done here, but hundreds or thousands of source-code files, figuring out flow of control between them, hopefully indicating if any code is missing, etc. Can you suggest?

#### 🕅 🕺 Thought for 21 seconds 🔨

Okay, the user wants to set up a local AI chatbot capable of analyzing code in C/C++, Java, and Python, handling large projects with thousands of files, and maintaining a chat interface similar to our conversation. They need this to work offline, so no internet connection. Let me think through how to approach this.

First, the core of the system would need a large language model (LLM) that's capable of code analysis. Since it's offline, they can't rely on cloud-based APIs like OpenAI or Gemini. So they need a locally deployable LLM. Models like LLaMA 2, CodeLlama, or Mistral might be suitable here. These models are open-source and can be run on local hardware. The challenge would be ensuring the model is fine-tuned for code understanding across the required languages.

Next, the system needs to handle large codebases. Traditional IDEs use static analysis tools to parse code, build abstract syntax trees (ASTs), and understand control flow. Integrating such tools would be essential. For C/C++, tools like Clang or ctags can generate ASTs. For Java, Eclipse JDT or JavaParser, and for Python, something like Jedi or Pyright. These tools can index the code, allowing the AI to query relationships between functions, variables, and files.

Then, there's the issue of scalability. Processing thousands of files requires efficient indexing and retrieval. A vector database like ChromaDB or FAISS could store embeddings of code snippets, enabling semantic search. This would help the AI find relevant code sections even if the user's query isn't exact. Additionally, a graph database like Neo4j could model the code structure, showing how different files and functions interconnect, which helps in tracking control flow and identifying missing dependencies.

The chat interface needs to support back-and-forth interactions. Frameworks like LangChain or Llamalndex can help integrate the LLM with the code analysis tools and databases. They allow chaining prompts, maintaining conversation history, and accessing external tools. For a GUI, something like Gradio or a custom web interface could provide the user interaction layer, all hosted locally.

Another consideration is the hardware. Running large models locally requires significant RAM and GPU resources. Quantized versions of models (like GGUF format) can reduce memory usage. The user might need a high-end workstation with a powerful GPU (e.g., NVIDIA RTX 4090) or multiple GPUs to handle the load, especially for large codebases.

Security is crucial since the system is air-gapped. All components must be open-source to avoid proprietary dependencies. Data should be encrypted at rest, and access controls implemented to prevent unauthorized access. Regular updates would need to be manually installed via secure channels.

Putting this all together, the user would set up a local server with the LLM, static analysis tools, and databases. The chat interface connects to this server, allowing the user to query the codebase. The AI uses the static analysis tools to understand code structure and the LLM to generate insights, with the databases aiding in quick retrieval of relevant information.

Potential challenges include ensuring the LLM's accuracy for code-specific tasks, handling very large codebases efficiently, and maintaining a user-friendly interface. Fine-tuning the model on code-related datasets and integrating robust error handling would be necessary. Also, the initial setup might be complex for non-technical users, so detailed documentation and scripts would help.

In summary, the solution involves deploying a local LLM, integrating static analysis tools, using databases for efficient code retrieval, and creating a chat interface that ties everything together, all while ensuring security and scalability.

For a **local, air-gapped system** that can perform in-depth code analysis across languages (C/C++, Java, Python) and handle large projects, here's a suggested architecture:

# **Core Components**

#### 1. Local LLM (Large Language Model)

- Model Options:
  - CodeLlama 34B/70B (specialized for code understanding).
  - StarCoder (trained on 80+ programming languages).
  - Mistral 7B/22B (general-purpose but code-capable).
  - WizardCoder (fine-tuned for code generation/analysis).
- Quantization: Use GGUF or GPTQ formats to reduce memory usage (e.g., 4-bit quantization).
- Tools:
  - Ilama.cpp or Ollama (local inference engines).
  - MLC LLM (for GPU/CPU optimization).

## 2. Code Analysis Backend

#### • Static Analysis Tools:

- Clang/LLVM (for C/C++ AST parsing).
- Eclipse JDT (Java AST and control flow).
- Pyright or Jedi (Python type checking and navigation).
- Project Indexing:
  - ctags/universal-ctags (symbol tagging across files).
  - Kythe or SourceGraph (cross-repo code navigation).

## 3. Vector/Graph Database

- ChromaDB or FAISS (vector storage for code embeddings).
- Neo4j (graph database for tracking code dependencies/control flow).

# 4. Chat Interface

- Local Chat UI:
  - Text Generation WebUI (supports local models and extensions).
  - LM Studio (user-friendly local LLM interface).
- Custom Integration:
  - Use LangChain or LlamaIndex to connect the LLM to code analysis tools.

## Workflow for Large Projects

# 1. Code Ingestion:

- $\circ~$  Index the entire codebase with  ${\bf ctags}~{\rm or}~{\bf Kythe}$  to map functions, variables, and dependencies.
- Generate ASTs for critical files using language-specific tools (e.g., Clang for C/C++).

#### 2. Control Flow Analysis:

- Use Neo4j to visualize call graphs and data flow (e.g., "Where is function\_X called?").
- Flag missing dependencies or orphaned files via graph traversal.

## 3. Chat-Driven Queries:

- Ask questions like:
  - "How does data flow from moduleA.py to moduleB.java?"
  - "Is there error handling for this C function in the codebase?"
- The LLM parses indexed data and static analysis results to generate answers.

# 4. Anomaly Detection:

 $\circ\;$  Use the LLM to flag inconsistencies (e.g., unhandled edge cases, dead code).

# Hardware Requirements

- Minimum:
  - 64GB RAM, 12+ CPU cores, NVIDIA RTX 4090 (24GB VRAM) for GPU acceleration.
- Ideal:
  - Multi-GPU setup (e.g., 2x A100 80GB) for large codebases.
  - SSD storage for fast indexing.

# Security & Air-Gapped Setup

- Isolation:
  - Run all tools in Docker containers with no external network access.
  - Use Vault or KeepassXC for local secrets management.
- Updates:
  - Manually transfer model weights/indices via encrypted USB drives.

# **Open-Source Tools to Combine**

Category	Tools
Local LLM	CodeLlama, StarCoder, Ollama, llama.cpp
Code Analysis	Clang, Eclipse JDT, Pyright, Kythe, SourceGraph
Vector/Graph DB	ChromaDB, Neo4j, Apache Age
υι/υχ	Text Generation WebUI, JupyterLab, VS Code (with local extensions)

# **Example Use Case**

- Task: "Find all security vulnerabilities in this C++ project."
- Steps:
  - 1. The LLM queries the Clang AST for unsafe functions ( strcpy , malloc ).
  - 2. Cross-references with Neo4j to find callers of these functions.
  - 3. Flags unvalidated inputs using control-flow graphs.
  - 4. Generates a report via the chat interface.

# Challenges & Mitigations

- Scale:
  - For massive codebases, precompute embeddings/indexes offline.
- Accuracy:
  - Fine-tune the LLM on your codebase for domain-specific terms.
- Missing Code:
  - Use **Kythe** or **ctags** to detect unresolved symbols.

This setup would let you replicate cloud-like code analysis locally while maintaining full control over security and data. Let me know if you need specifics on tool configurations!

6 C 🌢 🖓

Very helpful. How about yourself, DeepSeek? Is there a version that could run locally without requiring an internet connection (after it was installed and configured, of course)? This is important for source-code examination in intellectual property litigation, where source code is often examined in locked-down environments under a Court Protective Order.

#### 🕅 🕅 🕅 🕅 🕅 🕅

Okay, the user is asking about whether there's a local version of DeepSeek that can run without an internet connection, especially for use in legal environments like intellectual property litigation where code needs to be examined under strict security.

First, I need to recall what I know about DeepSeek. From previous knowledge, DeepSeek offers various AI models, some of which are cloud-based. But the user is specifically asking about a local deployment. I remember that many AI models, like CodeLLama or Mistral, have versions that can be run locally using tools like llama.cpp or Ollama. So, maybe DeepSeek has similar offerings?

I should check if DeepSeek provides on-premises solutions or local deployments. If they do, I need to outline the requirements, like hardware specs, installation steps, and integration with other tools. If not, suggest alternatives that can be used locally, such as CodeLLama or StarCoder, and how they can be set up in a secure environment.

Also, the user mentioned court protective orders, so security is paramount. I should emphasize encryption, air-gapped systems, and auditing capabilities. Maybe mention Docker containers for isolation, secure data handling, and compliance with legal standards.

I need to structure the answer clearly, starting with whether DeepSeek offers a local version, then alternatives if not, and best practices for secure deployment. Make sure to cover hardware, software, integration with static analysis tools, and security measures.

Wait, the user also asked about handling large codebases with hundreds or thousands of files. So, the solution should include tools for indexing and navigating large projects, like ctags, Kythe, or SourceGraph. Maybe mention how these integrate with local LLMs.

Additionally, the user wants a chat-like interface for back-and-forth interaction. Tools like Text Generation WebUI or LM Studio could be suggested for a user-friendly interface. Also, using LangChain or LlamaIndex to connect the LLM with code analysis tools.

I should also address the importance of reproducibility and audit trails in legal contexts. Logging all interactions and analyses to ensure accountability.

If DeepSeek doesn't have a local version, it's important to be honest but provide viable alternatives. Maybe even if they don't, there might be ways to use their models via APIs in a controlled environment, but that would require internet, which isn't suitable here. So stick to fully local solutions.

Finally, summarize the recommendations, highlighting key tools and practices to meet the user's needs for security, local processing, and comprehensive code analysis.

For **intellectual property litigation** and other scenarios requiring air-gapped, secure code analysis, here's how to approach local deployment of AI tools like DeepSeek, along with alternatives if a fully offline version isn't available:

# **Current Status of DeepSeek for Local Deployment**

As of my knowledge cutoff (July 2024), **DeepSeek does not officially offer a self-contained, fully offline version** of its models for local deployment. However, there are **alternatives and workarounds** to replicate its capabilities in a locked-down environment:

# **Option 1: Local Alternatives to DeepSeek**

If you need a fully offline solution, consider these open-source/local LLMs specialized for code analysis:

Model	Use Case	Local Tools for Deployment
CodeLlama 70B	C/C++, Java, Python code analysis	llama.cpp, Ollama, MLC LLM
StarCoder	Multi-language code understanding	HuggingFace transformers
WizardCoder-34B	Code generation/analysis	text-generation-webui
Mistral 7B/22B	General code reasoning	LM Studio, llama.cpp

#### **Key Features:**

- Air-Gapped Compliance: Run entirely offline via quantized models (GGUF/GPTQ formats).
- **Code-Specific Capabilities**: Trained on GitHub, Stack Overflow, etc., for code navigation, summarization, and vulnerability detection.
- Integration: Pair with static analysis tools (Clang, Pyright) for cross-file control-flow tracking.

#### **Option 2: On-Premises Enterprise LLMs**

If DeepSeek is required, inquire about **enterprise on-premises licensing** (common for litigation-sensitive use cases):

- 1. Custom Deployment: Work with DeepSeek's enterprise team to deploy models on your hardware.
- 2. Hardware Requirements: Typically requires NVIDIA GPUs (A100/H100) with vGPU isolation.
- 3. Security: Ensure FIPS 140-2 compliance, full disk encryption, and audit logging.

# Workflow for Secure Code Analysis

#### 1. System Setup:

- Hardware: Use a dedicated workstation with NVIDIA RTX 6000 Ada (48GB VRAM) or similar for large models.
- Isolation: Run tools in Docker containers with no internet access (e.g., docker run --network none).
- Data Handling: Encrypt codebases with VeraCrypt or LUKS before ingestion.

## 2. Toolchain Integration:

- Static Analysis:
  - C/C++: Clang + CodeQL for AST generation and taint tracking.
  - Java: Eclipse JDT Core for cross-file dependency graphs.
  - Python: Pyright + Bandit for type checking and security flaws.

- Indexing: Use ctags or Kythe to map relationships across thousands of files.
- LLM Interface: Use Text Generation WebUI or JupyterLab for chat-like Q/A.

# 3. Query Examples:

- "Trace all calls to encryptData() in this Java codebase."
- "Identify unlicensed third-party libraries in C/C++ includes."
- "Flag functions without error handling in Python modules."

# **Compliance with Court Protective Orders**

# 1. Audit Trails:

- Log all LLM interactions and code queries to **immutable storage** (e.g., Write-Once Read-Many drives).
- Use git-annex or IPFS for version-controlled analysis outputs.
- 2. Redaction: Tools like Apache PDFBox or Redactable to sanitize outputs before sharing.
- 3. Expert Validation: Pair LLM outputs with manual review by a technical expert (required for legal admissibility).

# Alternatives if DeepSeek Isn't Offline-Ready

#### 1. Precompute Analysis:

- Run DeepSeek via API in a secure, pre-approved environment, then export results to air-gapped systems.
- 2. Hybrid Workflow:
  - Use local LLMs (CodeLlama) for daily analysis, reserving DeepSeek for specific tasks in controlled sessions.

# **Final Recommendation**

For strict Court Protective Order compliance:

- 1. Deploy CodeLlama 70B or StarCoder via 11ama.cpp on an air-gapped machine.
- 2. Pair with Eclipse JDT (Java), Clang (C/C++), and Pyright (Python) for static analysis.
- 3. Use Neo4j or Apache Age to visualize cross-file dependencies and detect missing code.

This setup balances security, compliance, and technical depth while avoiding cloud dependencies. Let me know if you need specifics on hardening the OS or configuring GPU passthrough!

< 2/2 > 🕤 💭 խ 🖓

