

Episode AI: The Phantom Dependency Menace

What's in your AI Code?

 ENDOR LABS

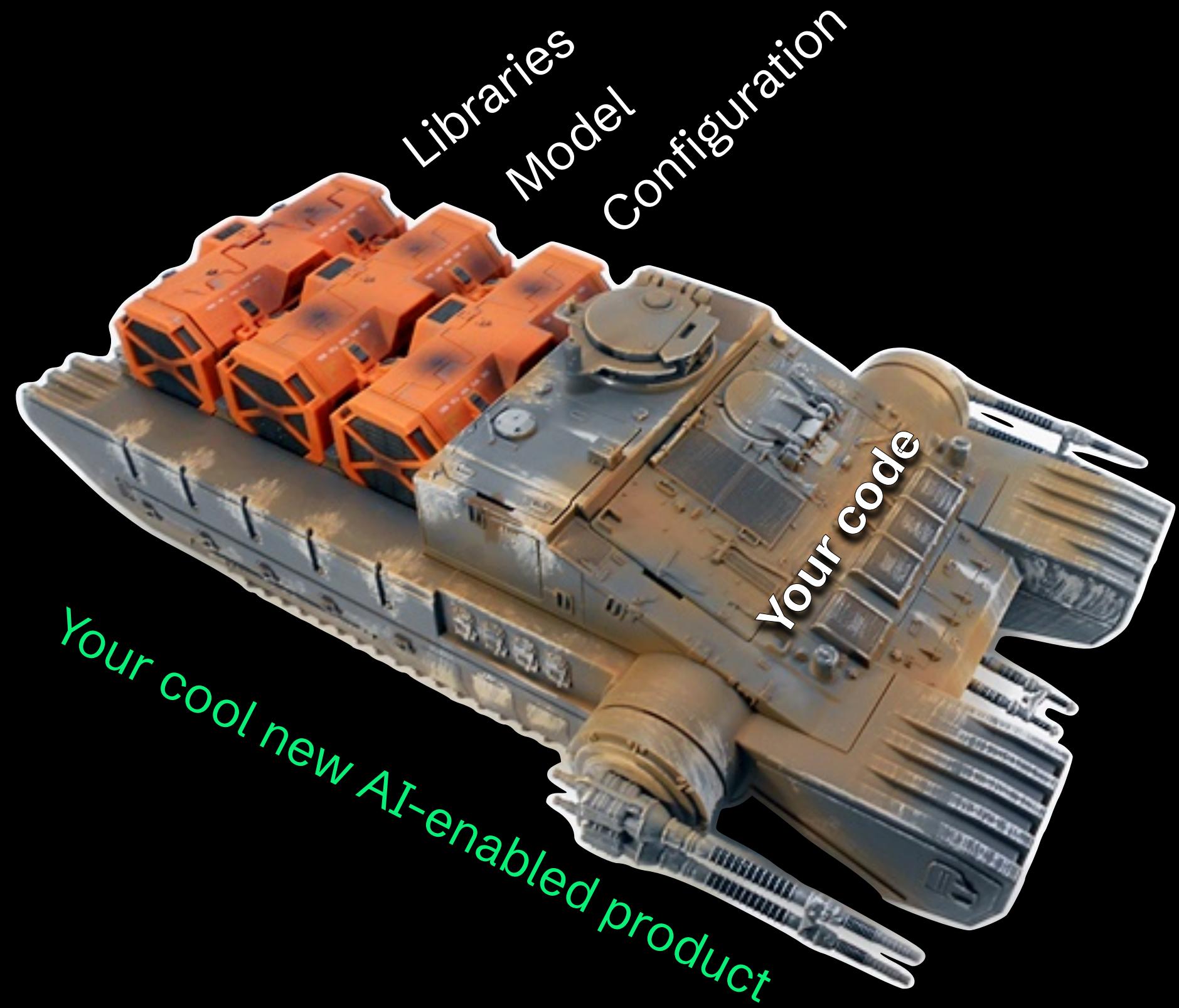
Darren Meyer, Staff Research Engineer – 22. October 2024

Why me?

- 20 years in AppSec
 - Built and ran security programs
 - Practical researcher for over 9 years
- Software engineering background
- General nerd



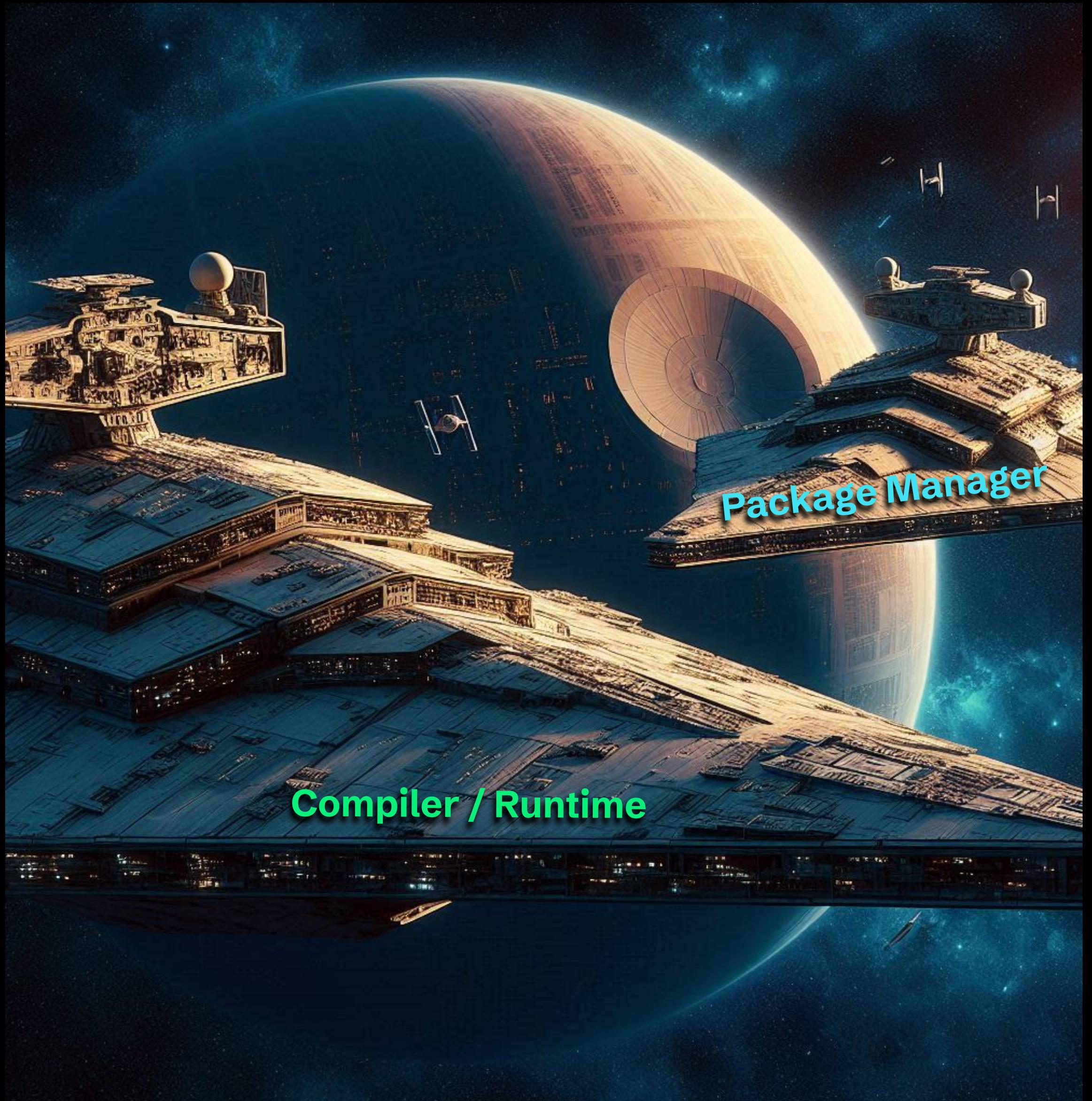
What do we mean by “AI Code”?



How dependencies work

Package Managers and Runtimes tend to operate completely decoupled, like ships passing in the night

1. Developer creates a **manifest** file (requirements.txt, package.json, pom.xml, etc.) to declare **direct** dependencies
2. Build system runs **package manager** and the direct dependencies bring along several other **transitive** dependencies
3. Package manager copies the files in a directory
4. **Runtime/compiler** loads dependencies as needed during execution



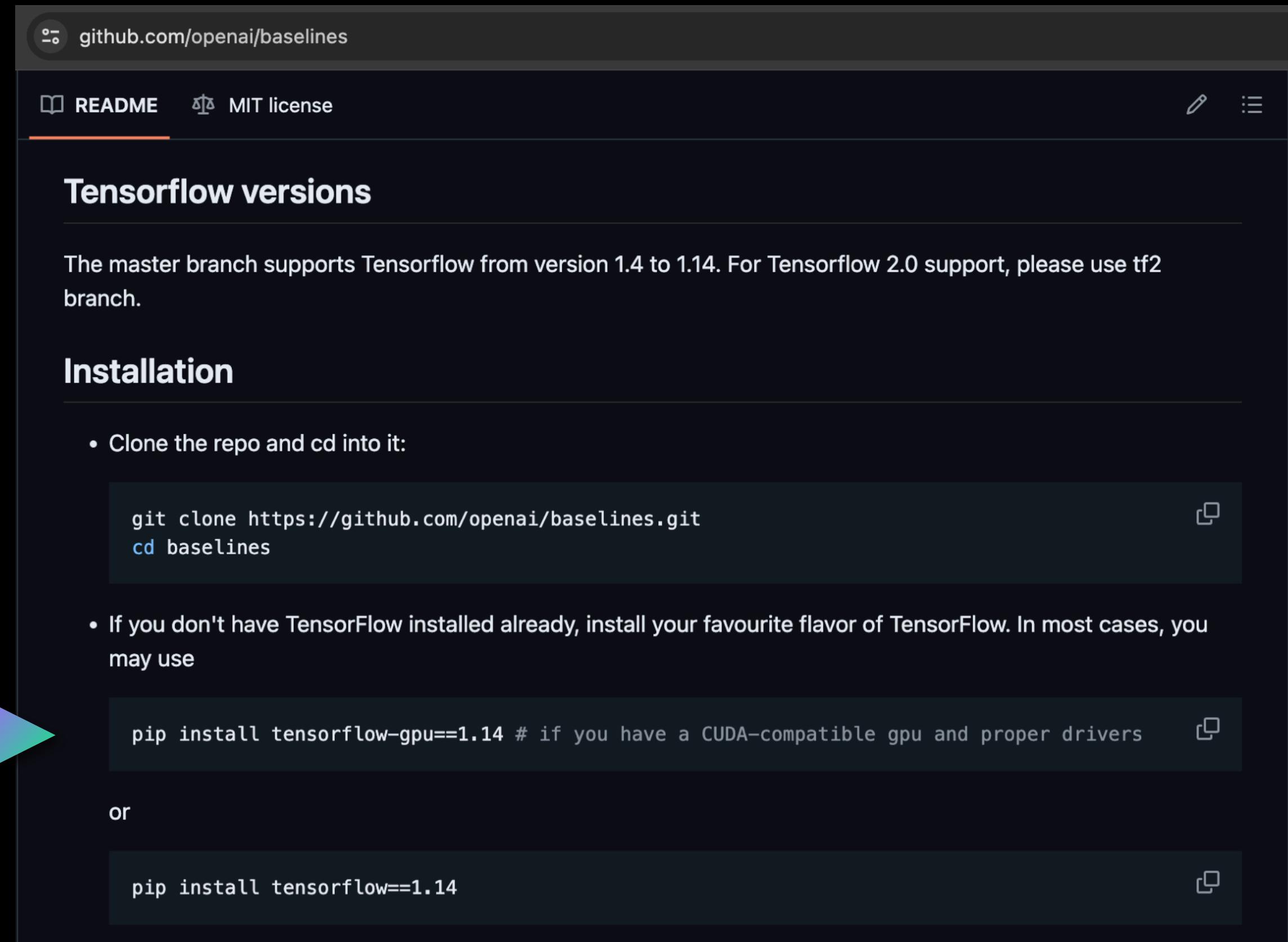


Things get out of sync

It is unavoidable

- Developers import new dependencies without updating the manifest file (possible in python, Javascript, scripts etc)
- In some cases dependencies are there in the environment (like global python or node packages)
- In some cases dependencies are for testing/dev but unmarked or misused
- In some cases dependencies are removed from the code but not from the package manager manifests

Your manifest can lie



The screenshot shows the GitHub README page for the `openai/baselines` repository. It includes sections for `Tensorflow versions`, `Installation`, and command-line installation instructions.

Tensorflow versions

The master branch supports Tensorflow from version 1.4 to 1.14. For Tensorflow 2.0 support, please use `tf2` branch.

Installation

- Clone the repo and cd into it:

```
git clone https://github.com/openai/baselines.git  
cd baselines
```
- If you don't have TensorFlow installed already, install your favourite flavor of TensorFlow. In most cases, you may use

```
pip install tensorflow-gpu==1.14 # if you have a CUDA-compatible gpu and proper drivers
```

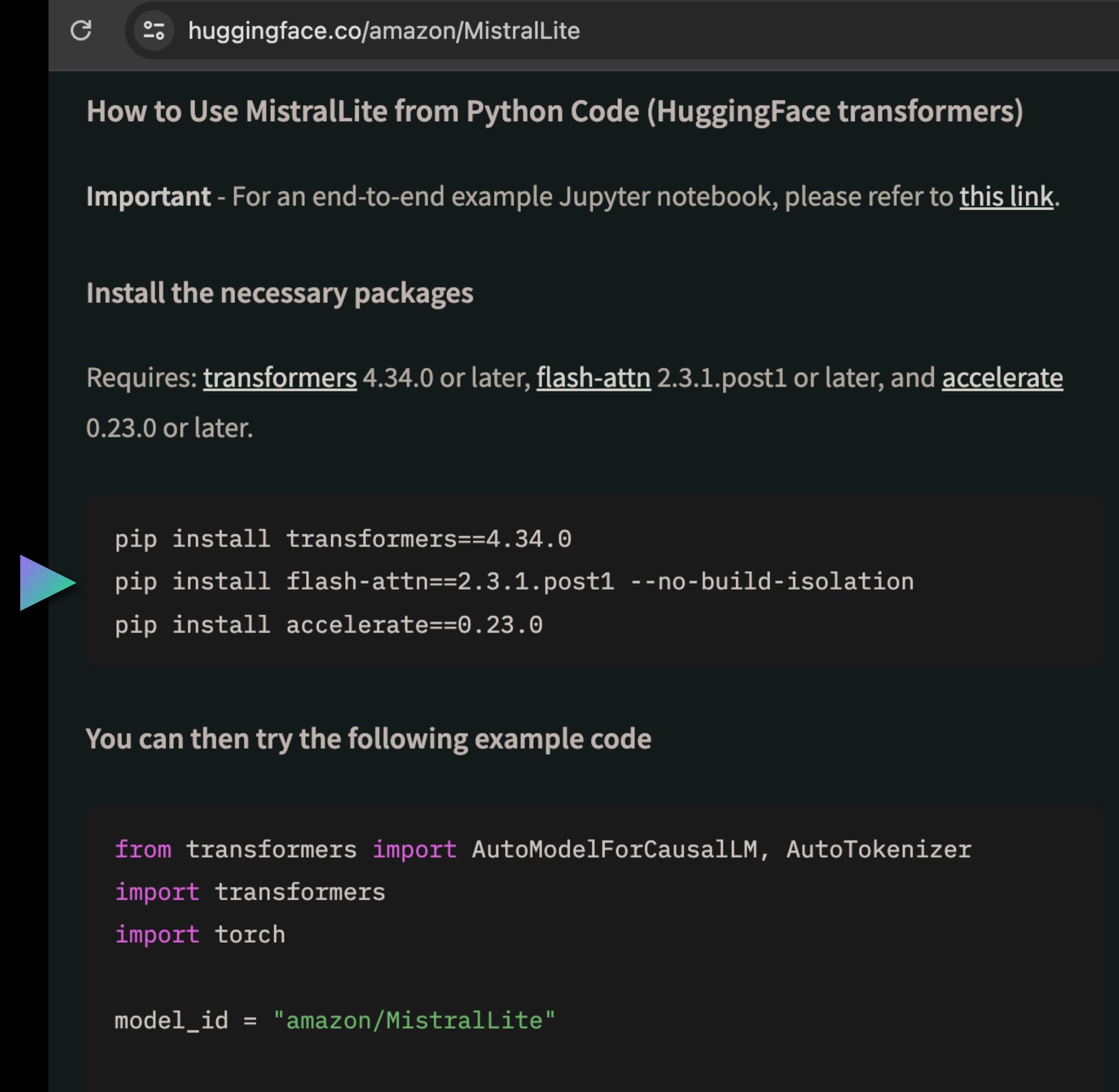
or

```
pip install tensorflow==1.14
```

OpenAI's Baselines library

- “Just `pip install` a dep”
- Baselines *won’t function without the right version*, but there are several “right versions”
- You’ll never see it in a manifest or lock file
- SCA / dep tree tools (usually) won’t see it

Models suggest this pattern often



The screenshot shows a browser window with the URL huggingface.co/amazon/MistralLite. The page title is "How to Use MistralLite from Python Code (HuggingFace transformers)". A note says "Important - For an end-to-end example Jupyter notebook, please refer to [this link](#)". Below it, a section titled "Install the necessary packages" lists requirements: "Requires: `transformers` 4.34.0 or later, `flash-attn` 2.3.1.post1 or later, and `accelerate` 0.23.0 or later." It includes a code block with three pip install commands:

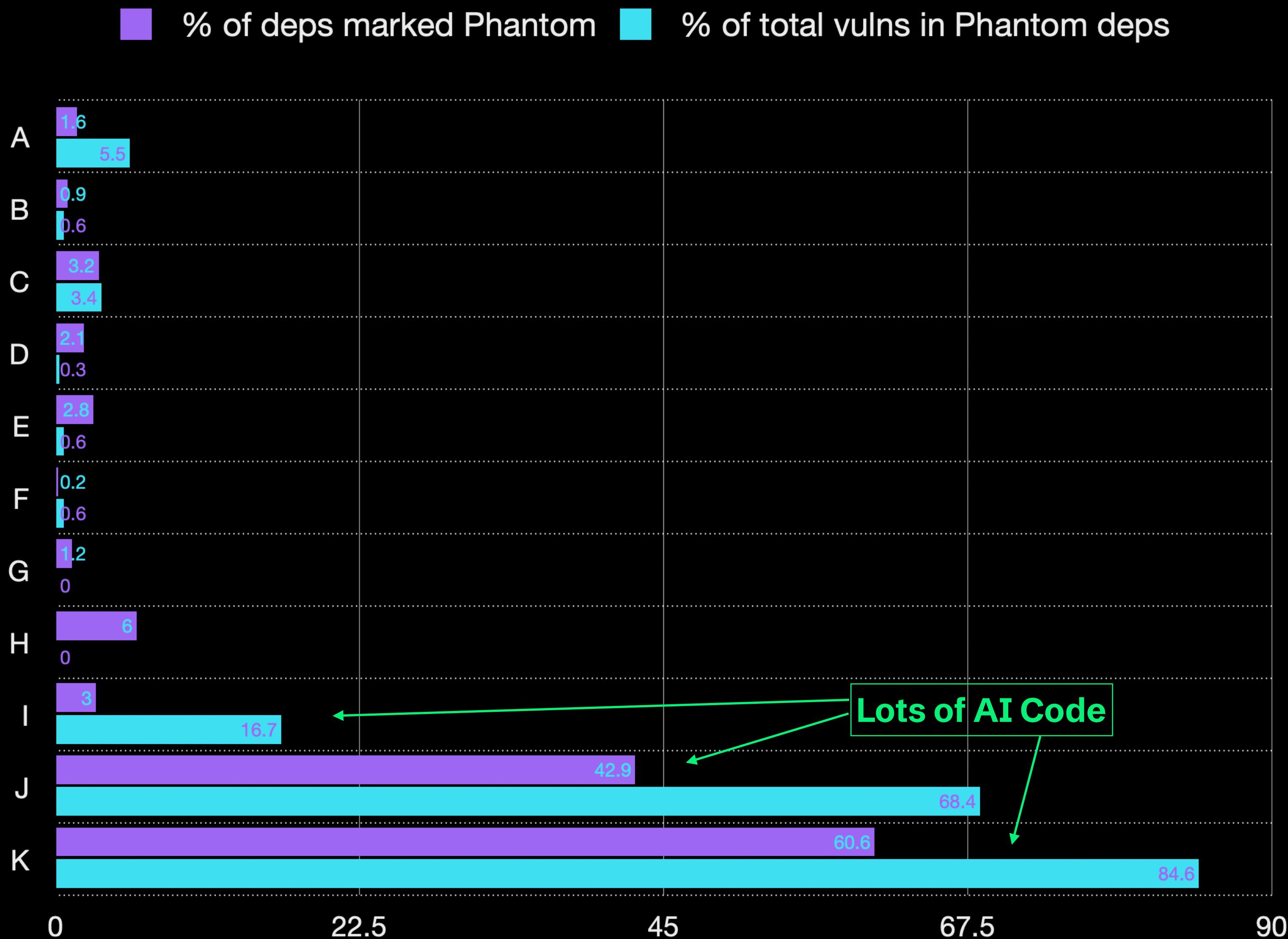
```
pip install transformers==4.34.0  
pip install flash-attn==2.3.1.post1 --no-build-isolation  
pip install accelerate==0.23.0
```

Below this, a section titled "You can then try the following example code" contains the following Python code:

```
from transformers import AutoModelForCausalLM, AutoTokenizer  
import transformers  
import torch  
  
model_id = "amazon/MistralLite"
```

The Phantom Dependency Menace

- Dependencies that are either “provided” by the system are assumed to be downloaded manually
- Scripts, containers, and so on
- Often depend on the target platform



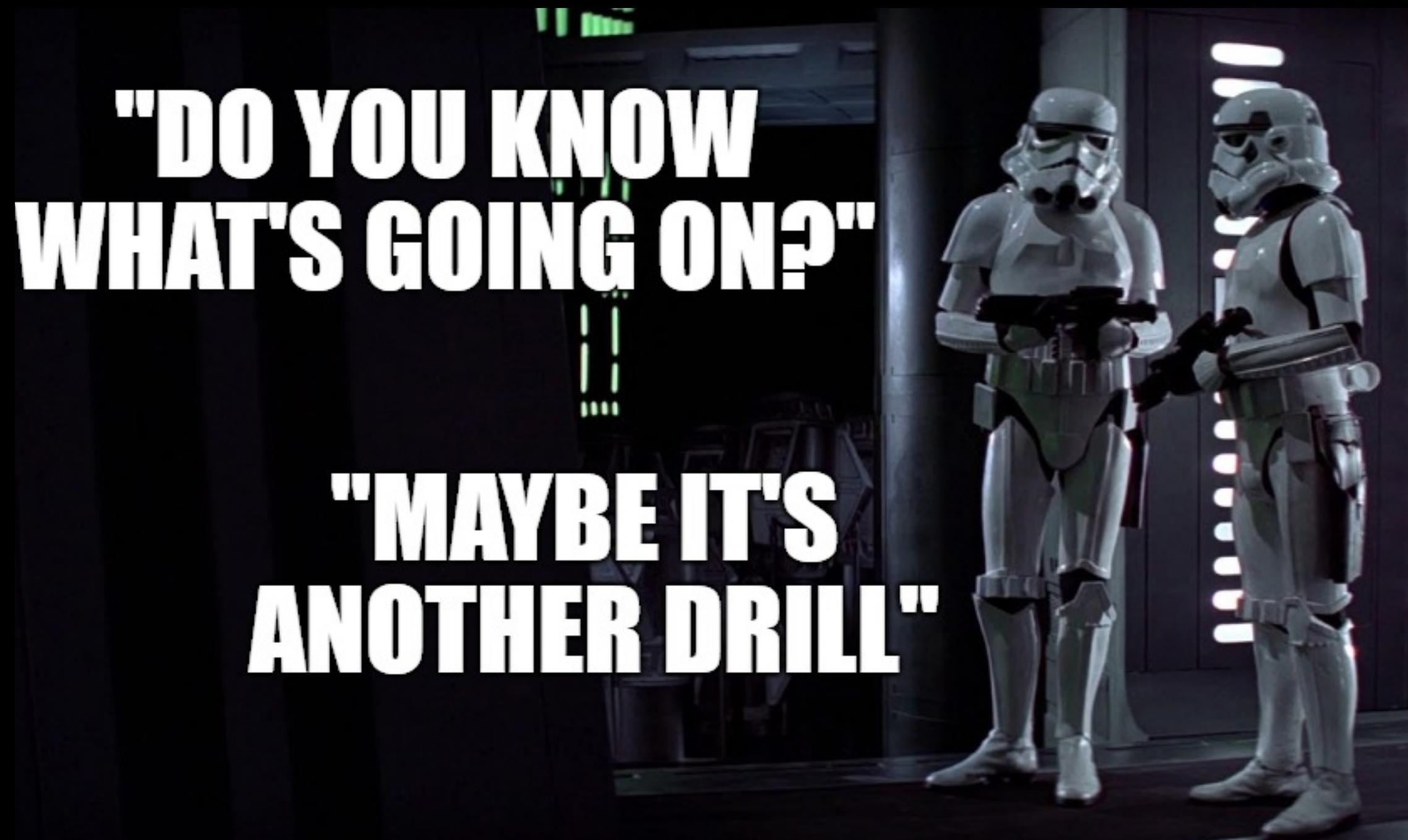
Security Challenges



- False sense of security — tools can't see what's not in a manifest, so you miss risks that might be relevant
- Inaccurate compliance data — your SBOMs aren't reporting everything in use. Auditors are unhappy if they catch you
- Dev / prod differences — can't rely on the version I see in dev pipelines being the same thing that's in production

Most tools are blind

Many tools trust the manifest or lock files, and don't account for the ways those can lie



1. Phantom Dependencies (false negatives)
 - Brought by the system, runtime or other scripts
2. Mis-used dependencies (false negatives)
 - Dependencies brought as “test/dev” used in runtime
3. Direct use of transitives (unreliable fixing)
 - Dependencies brought in as transitives and used directly without knowledge
4. Unused dependencies (false positives and noise)
 - Dependencies brought in the manifest but not used by the code

Program Analysis FTW

What matters is which packages the code **actually uses**

1. Source of truth is actually **the source code / bytecode**
 - a. Analyze the code
 - b. Create an Abstract Syntax Tree
 - c. Analyze types and call flows
 - d. Create a **call graph**
2. Correlate the **dependencies used by the code** with the dependencies fetch by the package manager or available in the file system
3. Create a unified view



Example: Python

Use the source

1. Import dependency ► Call graph ► “Is it used?”
 - a. Repeat for all it’s dependencies ([transitive](#))
 - b. Output: [dependency graph](#)
2. Compare [dependency graph](#) with versions installed on system and defined in manifest
3. [Correlate](#) and unify results
 - a. Makes accurate [SBOM](#) and [VEX](#) possible



Made possible with help from

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original research, program analysis design

Jamie Scott
models, reviews

Many contributors to OSS projects and papers



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