

Outline

- First, about the Al Alliance
- How non-deterministic GenAl affects testing
- What we can do about the challenges
- Adopt a new perspective?

This isn't a "problem solved!" talk. I'll describe the problem and outline potential solutions.

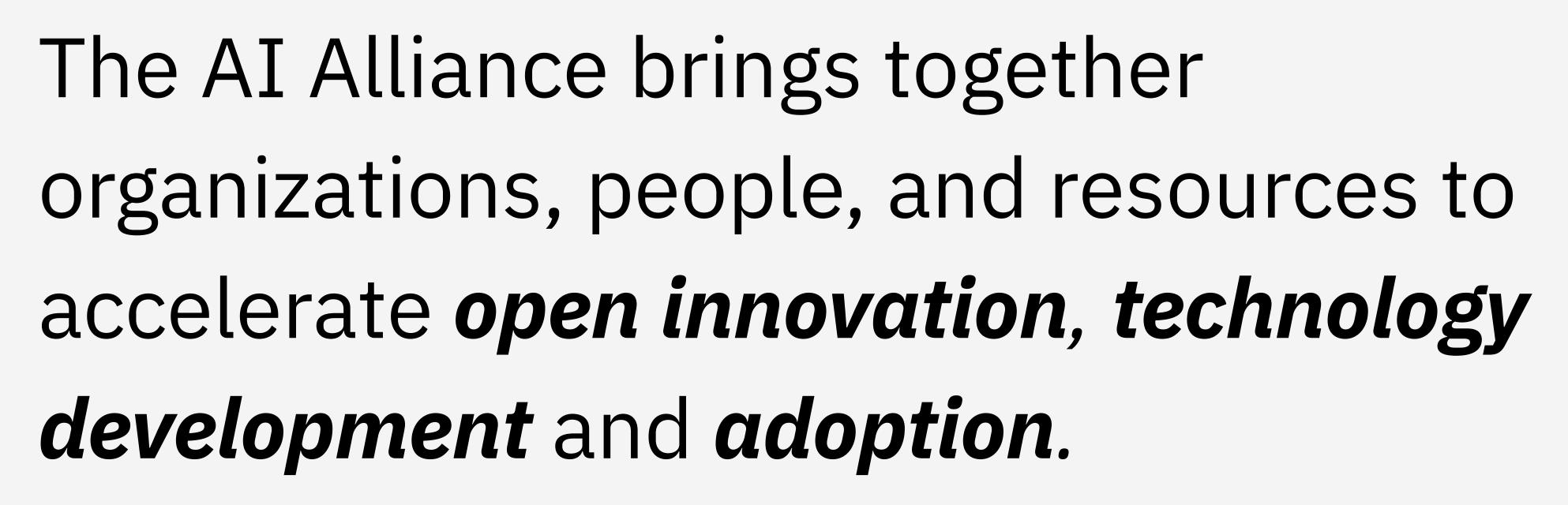


Outline

- First, about the Al Alliance
- How non-deterministic GenAl affects testing
- What we can do about the challenges
- Adopt a new perspective?

What's the connection to FP?
I'll discuss connections
as we go...





Launched December 5, 2023







Map of Members

Member organizations in the AI Alliance comprise academia, commercial, research and non-profits and span the globe.

Our core beliefs in AI that is open is the tie that binds us, despite our differences.

The AI Alliance is made up of +140 organizations in +20 countries, and growing





Focus Areas & Mission

Represents the investment priorities for the AI Alliance

1. Skills & Education 2. Trust & Safety

Support global AI skills building, education, and exploratory research.

Create benchmarks, tools, and methodologies to ensure and evaluate high-quality and safe AI.

3. Applications & Tools

Build and advance efficient and capable software frameworks for model builders and developers.

Member organizations have the choice to take part in one or more of these six focus areas and the agility to shift participation based on their interest and priorities.

4. HW Enablement

Foster a vibrant AI hardware accelerator ecosystem through SW.

5. Foundation Models & Data

Enable an ecosystem of open foundation models and datasets for diverse modalities.

6. Advocacy

Advocate for regulatory policies that create a healthy open ecosystem for AI.



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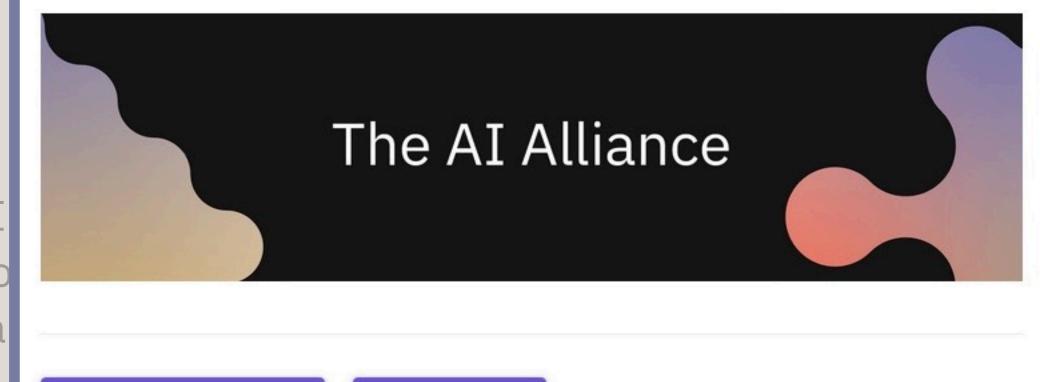
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Al Application Testing for Developers

GitHub Repo

: Applications and Tools ☐ (See the Contributors)
0.3, 2024-12-06
)

Tips:

Join Our Work Group

- 1 Use the search box at the top of this page to find specific content.
- 2 Capitalized, italicized terms link to a glossary of terms.

Welcome to the The Al Alliance project to advance the state of the art for Developer Testing for Generative Al ("GenAl") Applications.

Using nondeterministic, Genenerative AI Models in an application makes it difficult to write Deterministic, Repeatable, and Automatable tests. This is a serious concern for application developers, who are accustomed to and rely on determinism when they write Unit, Integration, and Acceptance tests to verify expected behavior and ensure that no Regressions occur as the application code base evolves.

What can be done about this problem?

3. Applications & Tools

Build and advance efficient and capable software frameworks for model builders and developers.

6. Advocacy

Advocate for regulatory policies that create a healthy open ecosystem for AI.

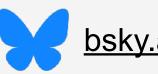


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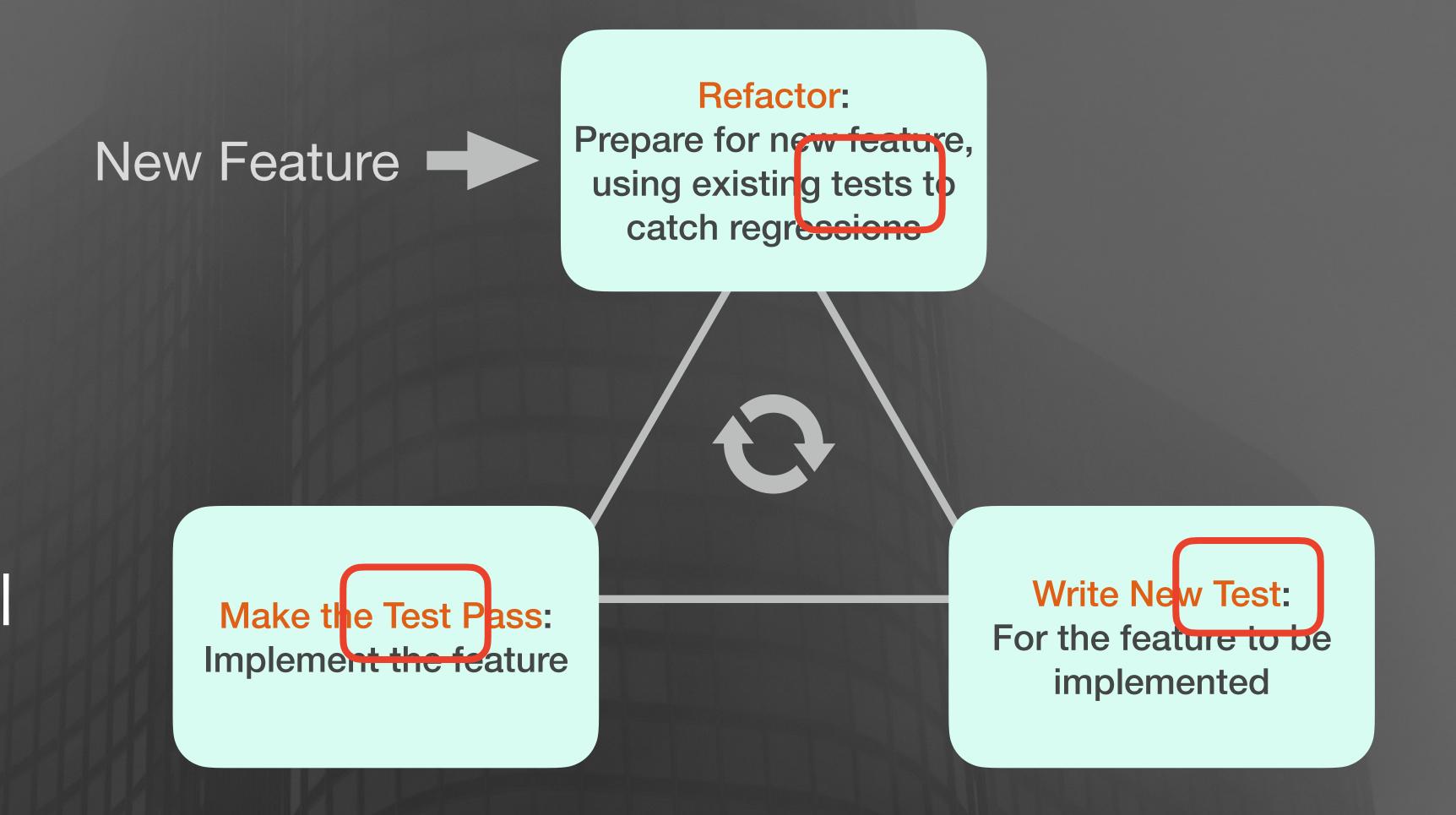








Remember the TDD loop?



Testing is integral to this process!



What Do Developers Expect?

Developers expect software to be deterministic[‡]:

- The same input \rightarrow the same output.
 - e.g., $sin(\pi) = -1$
- The output is different? Something is broken!
- Developers rely on determinism to help ensure correctness and reproducibility, and to catch regressions.



^{*} Distributed systems break this clean picture.

What Do Developers Expect?

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Put another way, the

• e.g., determinism makes it easier to

specify the system invariants.

What should remain true before and after each step? inistic[‡]:

roken! ensure atch

* Distributed systems break this clean picture.



What Do Developers Expect?

FP gave us property-based testing:

- E.g., QuickCheck, Hypothesis, ScalaCheck, ...
- Hypothesis example:

```
@given(st.integers(), nonzero_integers, st.integers(), nonzero_integers)
def test_two_non_identical_rationals_are_not_equal_to_each_other(self, numer1, denom1, numer2, denom2):
    Rule: a/b == c/d iff ad == bc
    This is a better test, because it randomly generates different instances.
    However, the test has to check for the case where the two values happen to be equivalent!
    """
    rat1 = Rational(numer1, denom1)
    rat2 = Rational(numer2, denom2)
    if numer1*denom2 == numer2*denom1:
        self.assertEqual(rat1, rat2)
    else:
        self.assertNotEqual(rat1, rat2)
```



What do we get with generative AI?

Generative models are probabilistic*:

- The same prompt → different output.
 - chatgpt("Write a poem") → insanity-

"Insanity is doing the same thing over and over again and expecting different results."

— not Einstein

* A tunable "temperature" controls how probabilistic.

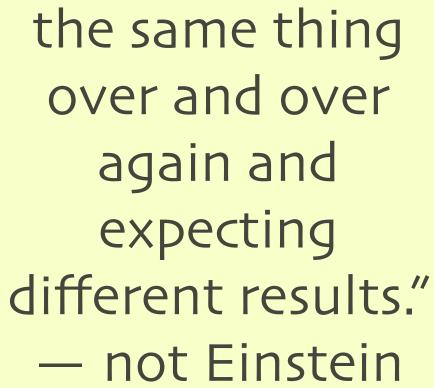


What do we get with generative AI?

Generative models are probabilistic*:

- The same prompt → different output.
 - chatgpt("Write a poem") → insanity
- Without determinism, how do you write repeatable, reliable tests? Specifically for GenAI,
 - Is that new model actually better or worse than the previous model, in my application?
 - Did any regressions in other behavior occur?

* A tunable "temperature" controls how probabilistic.



"Insanity is doing

What do we get with generative Al?

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 chail Put another way, the invariants are much less clear and therefore harder to define programmatically and enforce.

ite or GenAI, worse

* A tunable "temperature" controls how probabilistic.





What we can do about the challenges

- Don't forget about coupling and cohesion
- Use external tools for verification
- Adapt benchmarks "unit benchmarks"
- Use an LLM as a judge
- Understand and leverage statistics





- The non-deterministic AI model isn't the whole application. (E.g., <u>Agent</u> architectures)
 - Wrap the model in a good API.
 - Use deterministic test doubles for it.
 - Test everything else like you normally do.

- Writing a good API:
 - Engineer your prompts to constrain outputs.
 - Use tools like <u>Pydantic-Al</u> for type safety (<u>example</u>).
 - Select the Gen Al models that seem to work best with your tools.

- Writing
 - Engine
 - Use to (examp
 - Select best wi

Thinking about types encourages you to find ways to constrain model queries such that the responses are more closely aligned with your goals. to work

outputs.

ifety

- Writing
 - Engine
 - Use too
 (example)
 - Selectbest wi

Most Al-enabled apps won't be open-ended chatbots, but use Al to resiliently translate between human text and tool APIs, and translate tool-to-tool interactions, so we don't have to do that translation in code ourselves.

outputs.

ifety

to work

 However, tried and true C&C techniques don't help us test the model input and output behaviors themselves, nor do they eliminate the non-determinism that is unavoidable in our acceptance tests[‡].

* The integration tests that prove features are done.



- Are you asking a model to generate code?
 - Check it with a parser or compiler
 - Scan for security vulnerabilities
 - Check for excessive cyclomatic complexity
 - Check that only allowed third-party libraries and versions are used.

- Are you asking a model to generate code?
 - Using TDD? If you ask for code that makes your hand-written tests pass, does the generated code allow the tests to pass?
 - (Example)

- Are you
 - Usingyour hageneral
 - (Exar

Currently, I don't think many models are very good at generating powerful tests, but they can do a reasonable job generating code to pass the tests.

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ne

- Are you asking a model to do logic or reasoning?
 - Check it with a logic/reasoning engine
 - Or use that tool instead to create your logic!

- Are you asking a model to do planning?
 - Check it with a planning engine
 - Or use that tool instead to create your plan!

- Are you asking a model to generate possible chemicals or physical processes?
 - Try creating and testing the chemical in a lab.
 - Test the physical process with a simulator.
 - (Letting AI generate the "idea", then testing in a simulator may be cheaper than using the simulator to generate possible ideas.)





Adapt benchmarks - "unit benchmarks"

- Models are evaluated with benchmarks.
 - Use a large number of examples.
 - Typically cover a broad topic,
 - e.g., effective Q&A, detect hate speech, detect bias, measure throughput, ...
 - Return a single measurement, usually o-100%.



Adapt benchmarks - "unit benchmarks"

- Not the same thing as a developer "test".
- But can we adapt the idea for testing?
 - Use a very narrow scope.
 - Still use a lot of examples for higher confidence.
 - Return a single measurement, usually o-100%.
 - But at what threshold do you "pass"??



Adapt benchmarks - "unit benchmarks"

- Example: SQL queries generated from text.
 - Build a Q&A dataset that uses logged queries (expected answers) with appropriate human prompts (the queries).
 - Each unit benchmark might focus on one specific kind of common query.

This is also an example of using a model to translate between text and an "API".





Use an LLM as a judge

- You have probably chosen a small model for production, because it costs less to use.
- Use a bigger, smarter for test runs to "judge" responses.
 - You'll call it less often, so the cost won't be as much of an issue.

Need data for your unit benchmarks? Use a big model to synthesize data!

Use an LLM as a judge

- It can work like this:
 - A test sends a query to the model or app.
 - The query and the response are sent to a larger model with the question, "Is this a good response for this query? Answer yes or no, and if no, provide an explanation."
 - Fail the test if the answer is no.
 - Use the explanation to debug.

Understand and leverage statistics

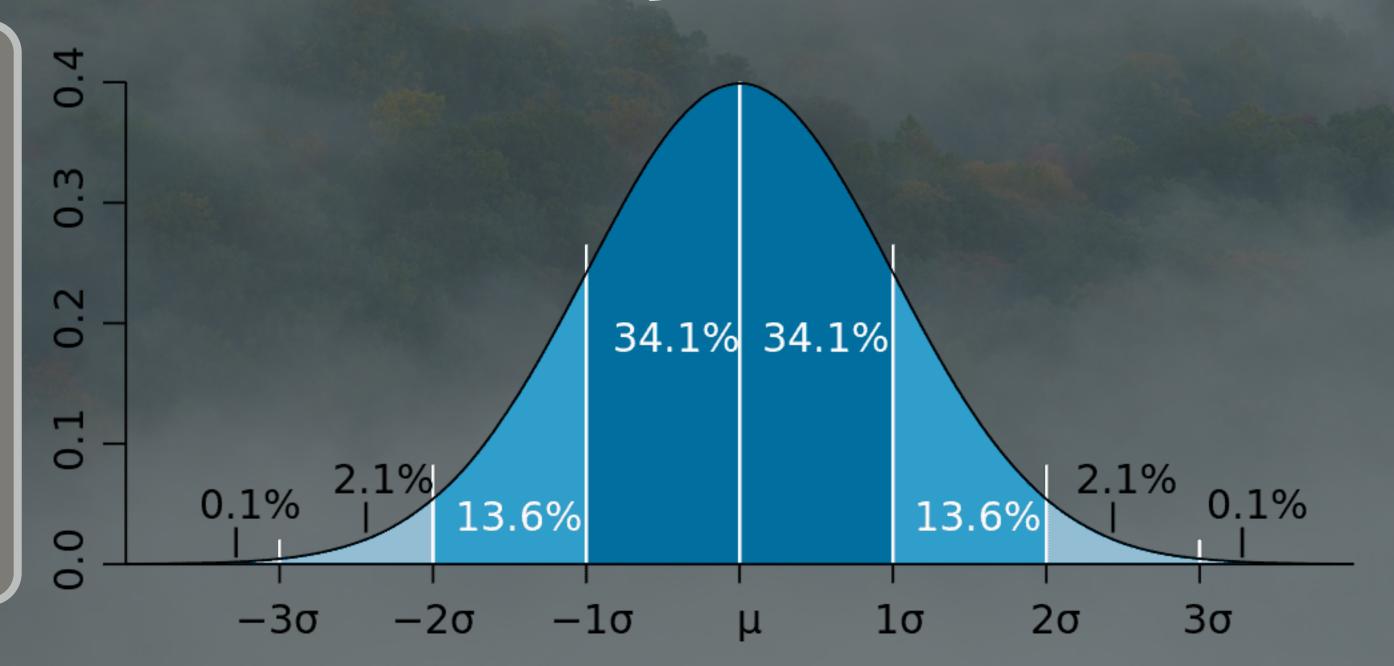


Understand and leverage statistics

- Scientists are accustomed to using statistics to analyze probabilistic phenomena.
 - E.g., a potential discrepancy between theory and experiment must be > <u>five sigma</u>.

"A five-sigma level translates to one chance in 3.5 million that a random fluctuation would yield the result."

Wikipedia



Understand and leverage statistics

- Classifier models sometimes return a confidence level, i.e., how much they believe they are returning the correct classification.
 - "Adding Error Bars to Evals: A Statistical Approach to Language Model Evaluations"
 - https://arxiv.org/abs/2411.00640





Anew perspective?

- The Structure of Scientific Revolutions
 - It's normal to try to bend our current theory to accommodate new data, rather than simply throw out our current theory and start over from the fundamentals.

Should we abandon the idea of deterministic testing, at least for model outputs, in favor of a new approach?



Anew perspective?

- ★What if we switch from verifying desired model behavior to coercing desired behavior, instead?
 - We already tune models to improve domainspecific knowledge, chatbot behavior, etc.
 - So,
 - Tired: Writing software and testing it.
 - Wired: Tune until satisfactory behavior is achieved.



Anew perspective?

 Changes to the TDD cycle (for model behaviors only):

New Feature



Refactor:

Prepare for new feature, using existing tests to catch regressions

Keep tuning until the unit benchmark passes

Make the Test Pass: Implement the feature Write New Test:

For the feature to be implemented

Write new unit benchmark

Thank you!

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* AI Alliance

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