This Year in Uber's Al-Driven Developer Productivity Revolution





Adam Huda Ty Smith



Uber's Scale



Annualized run-rate gross bookings



Monthly active platform consumers

10,000 + Cities



Trips per day



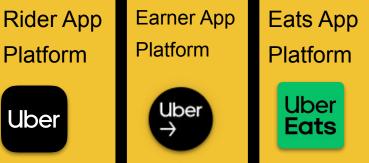
Monthly active drivers and couriers globally

70 Countries

Public figures for the quarter ended June 20, 2024

Platform Leverage

Feature Teams Backend, Frontend



Developer Platform Part of Platform Engineering

4,500 200

Feature engineers

Developer Platform engineers

22:1 Support ratio 5

Engineering sites

Developer Experience

IDEs & Tools

Training & Documentation

App & Service Frameworks

Internal Libraries & Guardrails

Monorepo Tooling & Build System



Monorepos for Swift, Kotlin, Typescript, Go, Java, and Python 5_{k+}

Microservices



6

Net promoter score

3

Major mobile apps Minor mobile apps

Technical Debt

Backlog of updates
Fragmentation
Test coverage

100_{million+}

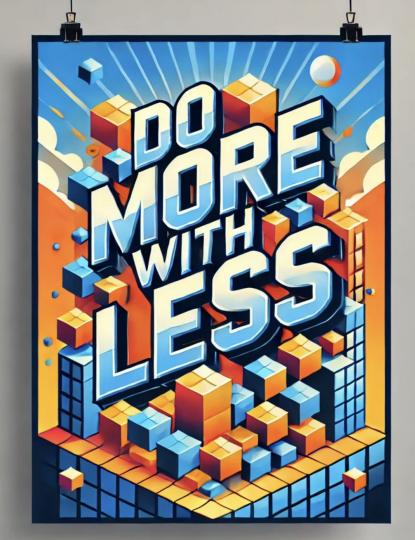
Lines of code across all of the monorepos



Macro Trends

Flat headcount

 Backfills not guaranteed



Emergence of Al as Leverage Can't scale people with the growth and maintenance needs of the codebase

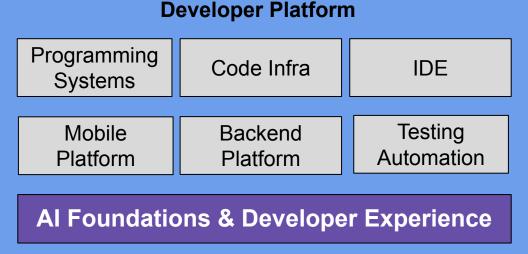
Can we position Developer Platform to be Al-driven?

Org High-level View

Created a centralized Al DevEx team that specializes in Al applied to the SDLC

Platform Engineering

Quality & Productivity Engineering



ML Infrastructure

Timeline Applied-Al Developer Tools



Inaugural Hackdayz Oct 2022

The ChatGPT moment Nov 2022

Exploring generative Al Hackdayz Summer 2023

Building automation and becoming Al-driven Hackdayz Winter 2024

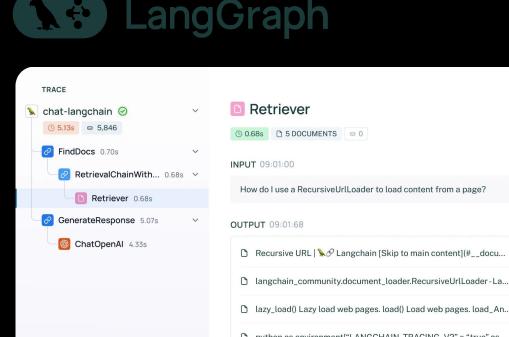
Focus on agentic systems Hackdayz Summer 2024

Agentic Systems

Multi-step systems to interact with LLMs

Breaks down a problem space into manageable tasks

You'll see some examples in our upcoming stories



LangChain

Applied-Al SDLC

DESTON CREVELOC S

Coding Assistants Generating Tests

RELEASE

OPERATE

3. Java to Kotlin Migration

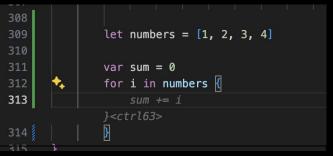


Code Assistants

Basics

- Native Plugin with Language
 Intelligence
- Model Backend

Contributes to UX and Result Quality



Broken Example Caused By IDE validation

IDE Plugin

Language Intelligence Semantics, syntax, references & symbol navigation

Verification Deterministic fixes for generated suggestions.

Extensibility Adapt to development environments, workflows, & custom tools

Model backend

Foundational model and context awareness

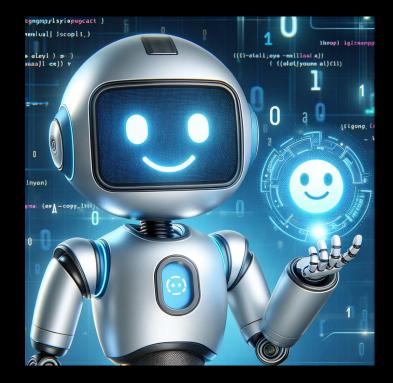
Hypothesis An Uber trained IDE assistant is needed

Custom Code Assistants

Requirements

- Uber aware
- Fast
- Cost effective
- Per user analytics
- Workflow integrated

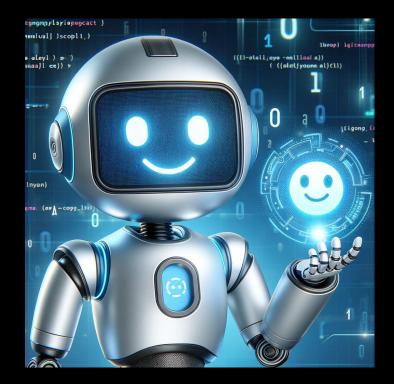




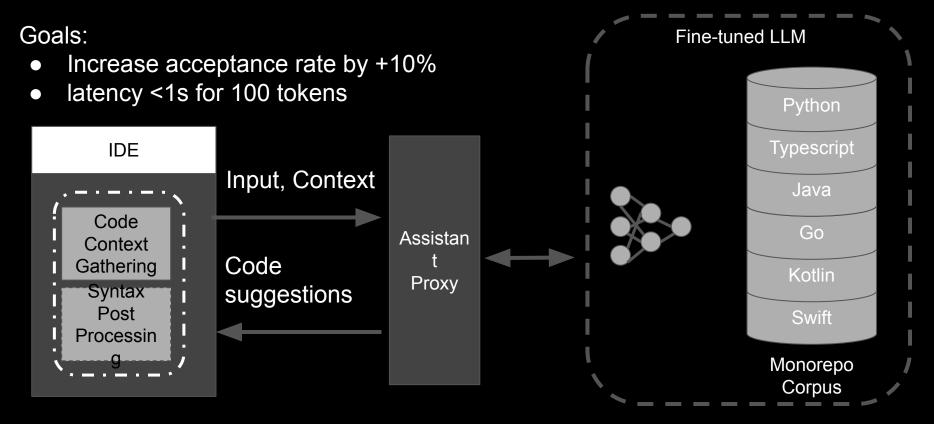
Custom Code Assistants

Buildout

- MVP in Hackathon
- Evaluate LLMs
- Internal evangelism
- Wide variety of investments



In-House Coding Assistant



Code Assistants

Downsides

- 6 months of work
- Underfunded
- Always playing catch up



What we learned

- MVPs are easy, productionisation is hard
- Latency requirements vary per tool
- User Experience matters
- UI surface cannibalization is a risk
- Follow ecosystem principle
- Continuously evaluate landscape



Focused on GitHub Copilot adoption & evangelism

Building on Industry Tools

Reusable Components

Code Context Gathering Summarize & rank code context to provide best input to use-cases:

Data Race Fixer	Linter Warning Flxer	
Crash Fixer		
`		

Gather telemetry

Assistan t Proxy Fine-tuned LLM

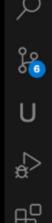


Custom model with knowledge of internal libraries, custom frameworks, and company-specific best practices

In-house Coding Assistant

Extend with chat @partcipants

@genie for monorepo knowledge base queries



 $\leftarrow \rightarrow$ ♀ go-code [SSH: shirhatti.devpod-us-or] . ப \Box @genie How do I mock tally? \times shirhatti_UBER ρ @genie How do I mock tally? 83 Genie To mock Tally in Go unit tests, you'll typically use a mocking framework like GoMock. H mocking Tally based on the context provided:

> 1. Add tallymock dependency to your test file: Import the necessary packages incl mocking package tallymock.

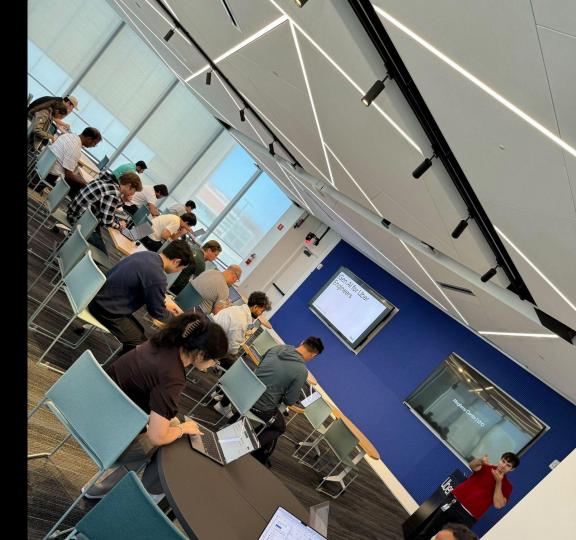
GitHub Copilot Coding Assistant

Hands-on workshops

Running once per month to spread knowledge

Experience iterating with LLMs and probabilistic outputs

GitHub Copilot Coding Assistant



Internal Evangelism Content

Chat participants	Chat commands	Chat context
@workspace @vscode	/doc /explain /fix	#codebase #editor #file
Extensible!	/tests …	#selection

A series of codelabs that teach these engineers about providing context, using chat participants, and commands to refactor code and generate tests

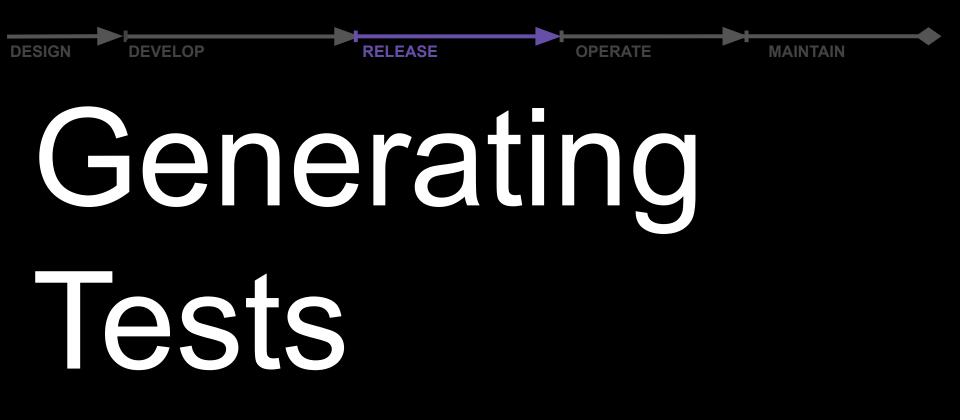
GitHub Copilot Coding Assistant

Coding Assistants

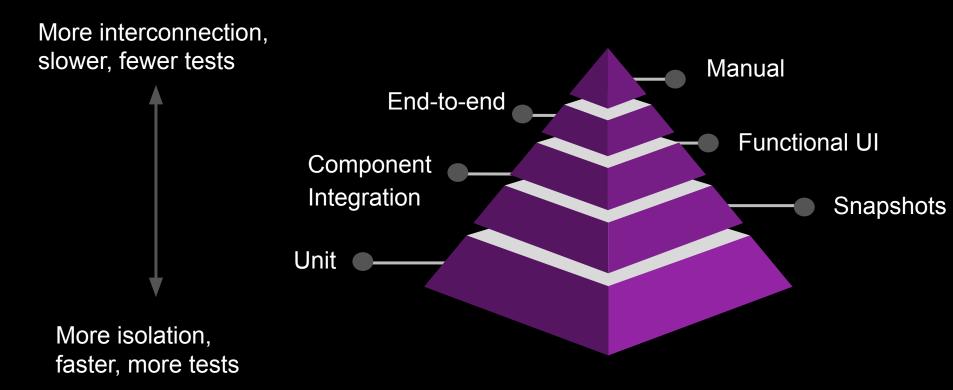
Future

- Platform Native
- Vendor Fine Tuning
- Extensibility Increases
- Open-source clients
- Multi Vendor Landscape
- Enterprise Requirements





Classic Testing Strategy



Testing Challenges

Writing good tests can be hard

Maintaining many different types of tests is tedious

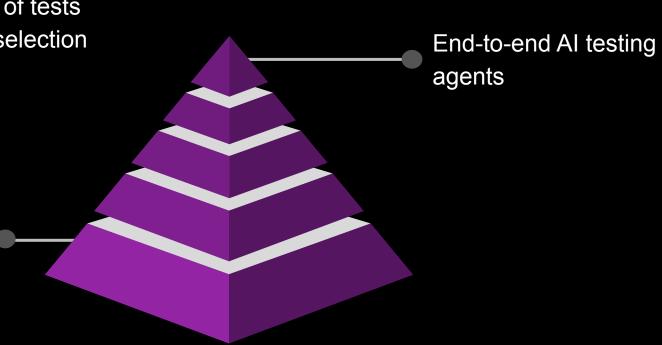
Permutations of languages, cities, experiments, platforms



AI-Driven Testing Pyramid

- Analyze quality of tests
- Predictive test selection

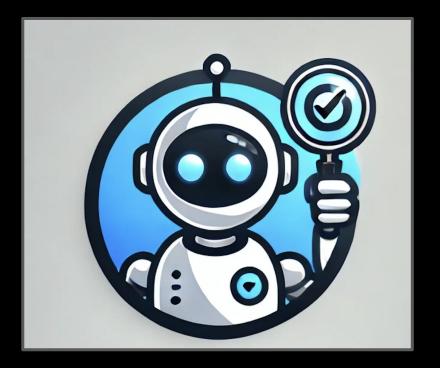
Al-powered code generation of unit tests



AutoCover Requirements

- Keep the developer-in-the loop
- Focus on regression tests

Increase coverage



AutoCover in action

Tests are streaming in.

+ 5 ... CHAT 🛞 GitHub Copilot Welcome, @matas_UBER, I'm your Copilot and I'm here to help you get things done faster. You can also start an inline chat session. I'm powered by AI, so surprises and mistakes are possible. Make sure to verify any generated code or suggestions, and share

feedback so that we can learn and improve. Check out the Copilot documentation to learn

the matas_UBER @autocover /start

8 Uber Auto Cover **Okay!** Generating tests

Ask Copilot 0 2 >~

more.

now.

co tasks.go

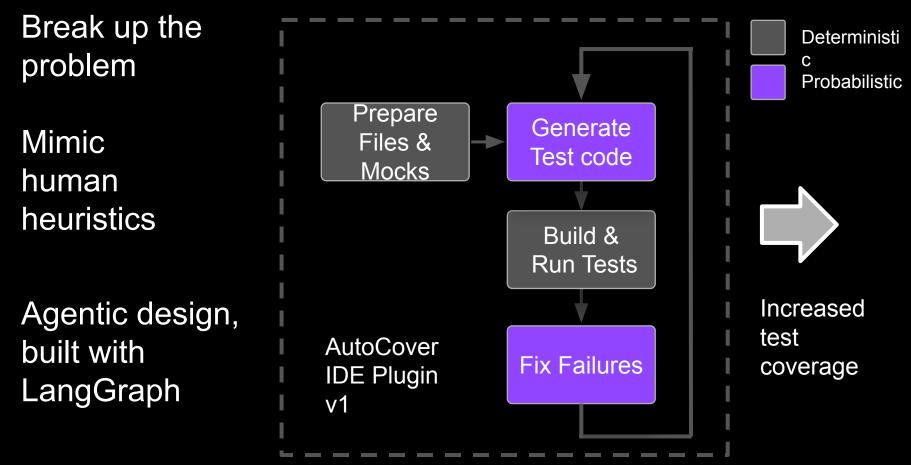
so tasks_test.go M X

src > code.uber.internal > crack > wallet > gateways > tasks > -co tasks_test.go You, 6 days ago | 1 author (You) package tasks 1

2

$\leftarrow \rightarrow$

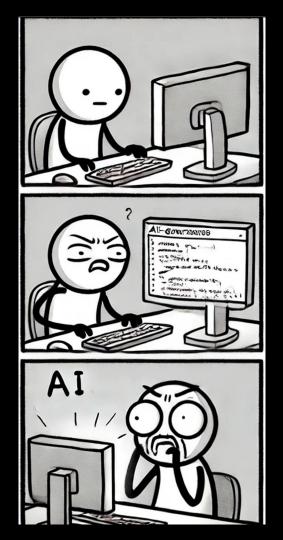
♀ go-code [SSH: matas.devpod-us-c



Generating Tests

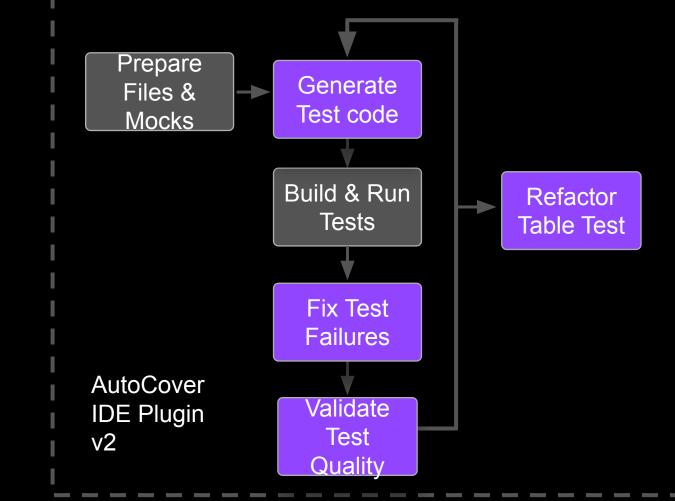
Automating away all this tedious work...

This needs to be labeled "generated with AI!"



Is it writing "change detector" tests? Validation step: Check assertions against intent

Refactor step: Adopt best practices like the table pattern



What's next

IDE/CLI for humans

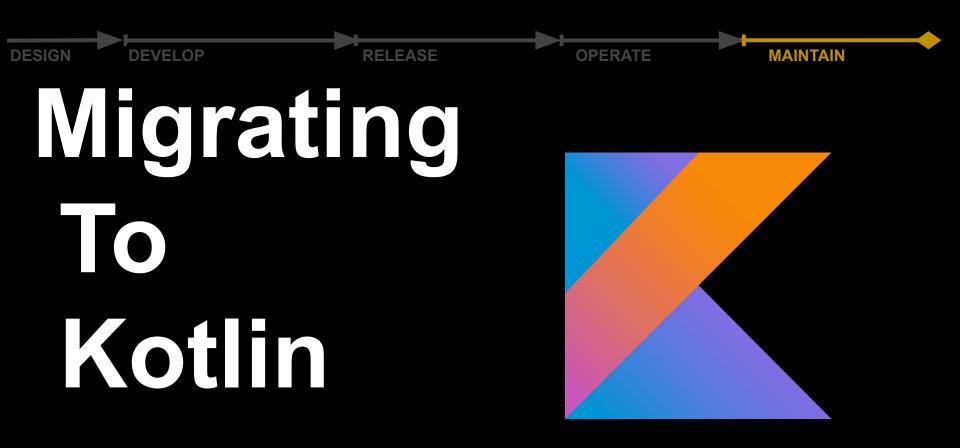
- Table test refactor
- Validate test quality

Headless mode

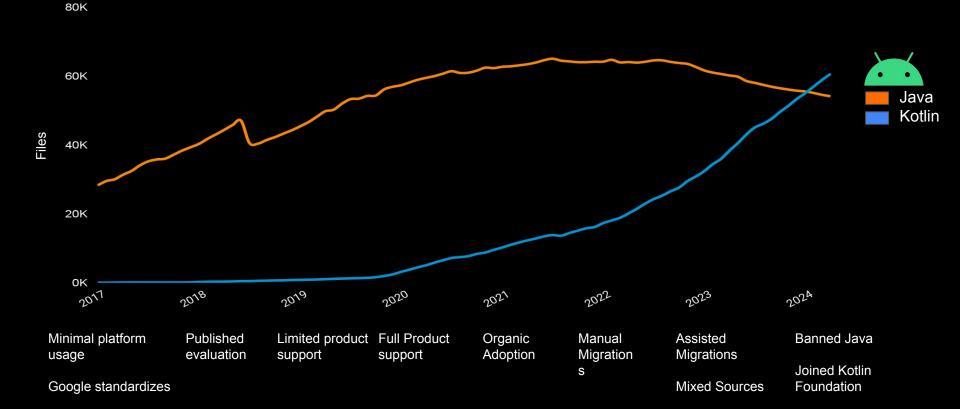
- Shift-right, runs on Cl
- Improve quality of existing test code

Mutation testing step

- Inject bugs ("mutants") into the source code
- See if it finds bugs
- Generate mutants with Al



Kotlin History at Uber



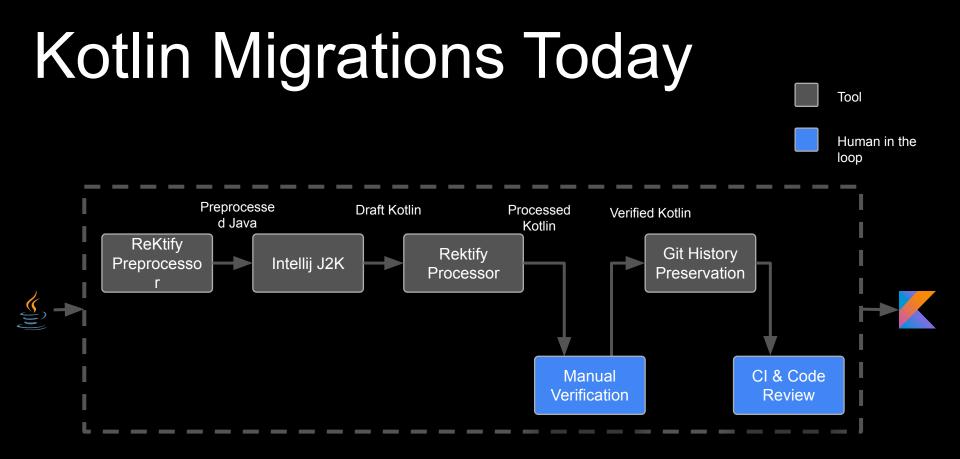
Kotlin Migrations

Decentralized

- Workflow incentives
- Industry standard
- Developer assisted

C Order	ReviewPluginFactoryTest.java ×	:
Modifying Java files is discouraged. Please use a mixed sources module & conve		Convert
	<pre>package com.ubercab.risk.action.open_order_review;</pre>	
	public class OrderReviewPluginFactoryTest extends UberTe	stBase {
	<pre>private OrderReviewPluginFactory pluginFactory;</pre>	
17	<pre>@Mock private OrderReviewPluginFactory.Parent parent;</pre>	

Uber J2K Conversion Steps
 1. Make RiskActionResultTracker ready for Kotlin conversion
 2. Convert RiskActionResultTracker from Java to Kotlin
 3. Fix the converted Kotlin code in RiskActionResultTracker with ReKtify
 4. Add History Aware Git Commits for RiskActionResultTracker
 Cancel OK



Rektify Processors

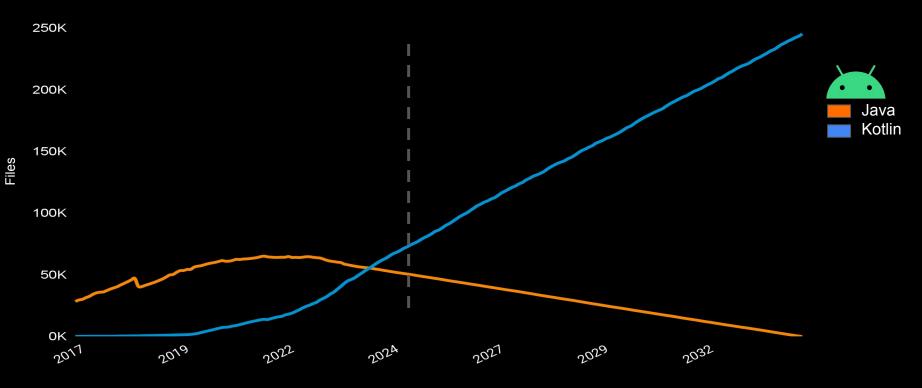
Pre Processors

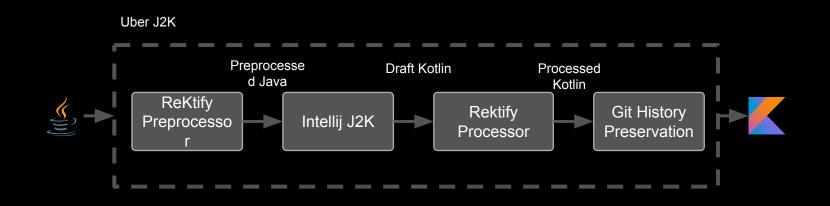
Nullable Annotations

Post Processors

- AndroidTextUtilsRule
- CaptorAnnotationRule
- GuavaStringUtilsRule
- LambdaExpressionRule
- MockAnnotationRule
- RemovelnitMocksRule
- AutoDisposeRule

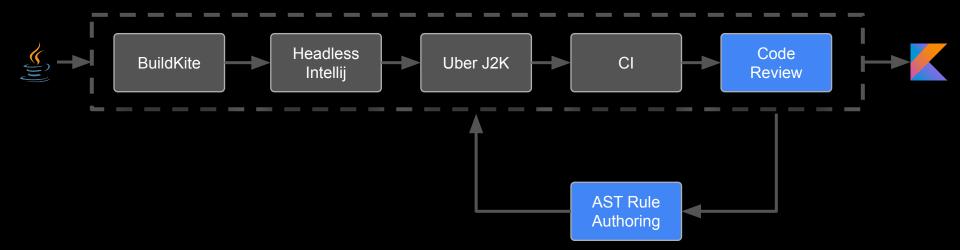
Kotlin Migrations



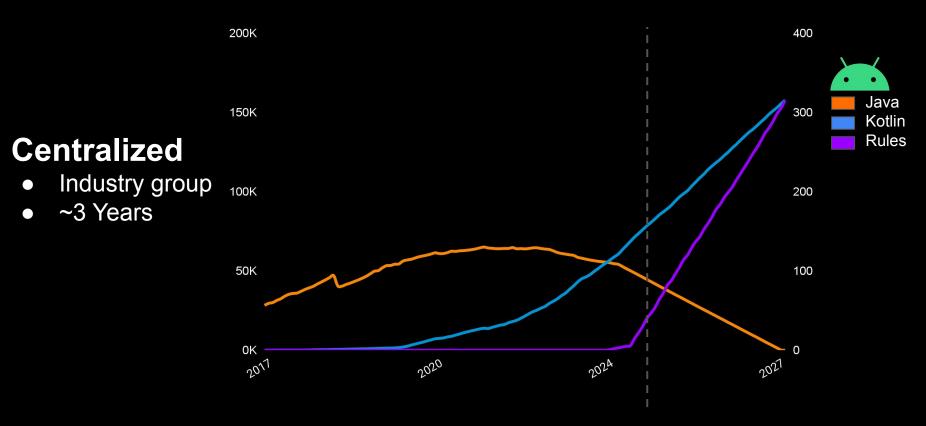








Kotlin Migrations



Can Al go faster?

LLM Kotlin Migration

Positives

- Flexible use cases
- Fast to deploy

Negatives

- Hallucinations possible
- High risk failures
- Humans are fallible

Combining AST + LLMs

AST + LLM

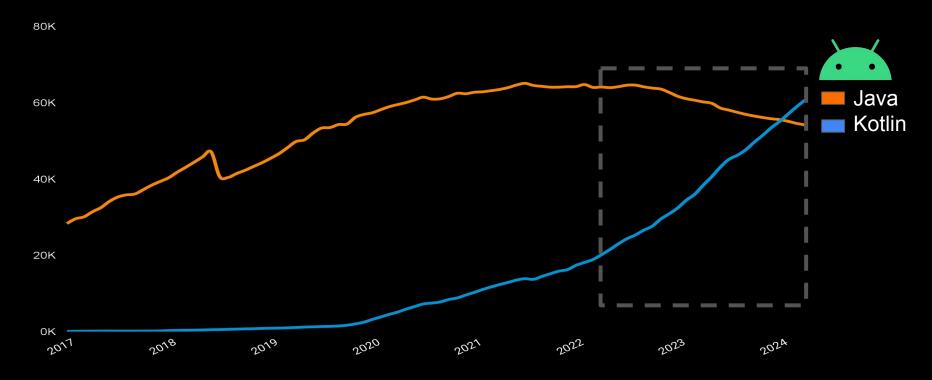
Positives

- Prior Art
- Deterministic
- Faster than human authored Rules

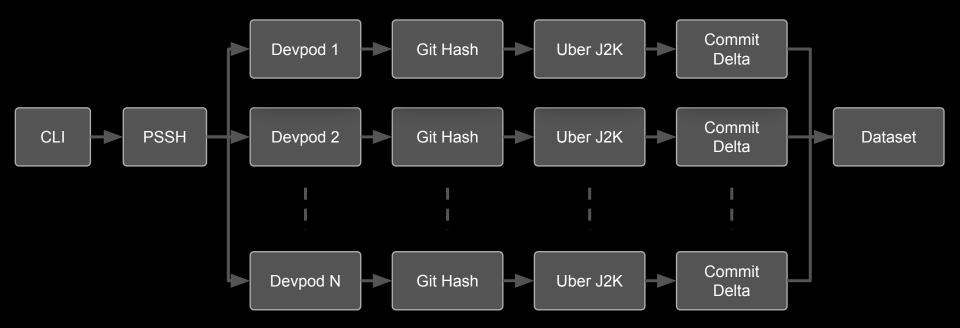
Negatives

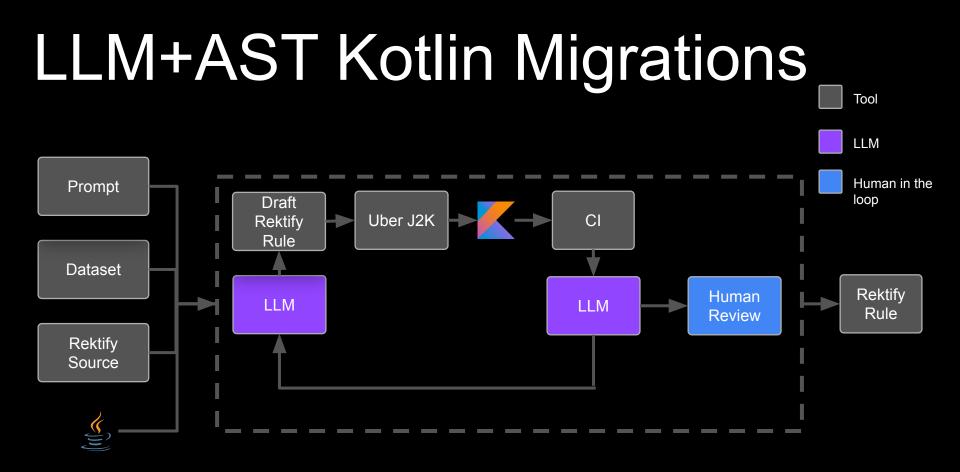
• Slower than LLM only

Dataset

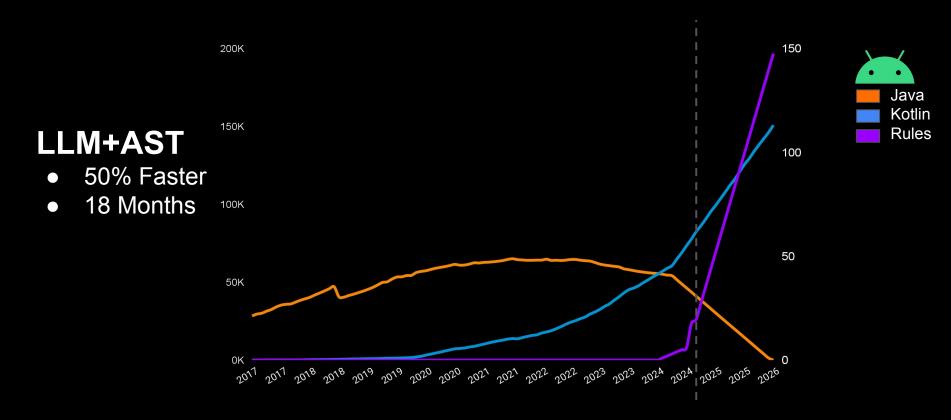


Dataset





Kotlin Migrations



Finishing the Migration

Challenges

- Rollout risk
- Begin in low risk areas
- Categorization of rules
- Noise fatigue

Questions

- LLM fallbacks
- Batch size
- Speed of innovation

Wrap up



Measuring AI impact

Challenges

- Fragmentation
- Organizational Cost

Risks

- Side Effects
- Wrong Investments
- Missed Investments

Measurement Philosophy

Lead with qualitative

Normalize quantitative impact on developer hours saved to prioritize bets



Developers report significant increase in productivity

1000+years

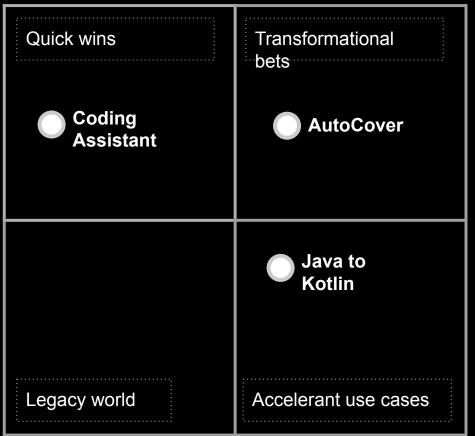
Potential impact of automating away of technical debt

Opportunities

Quality and velocity

Manage expectations and hype

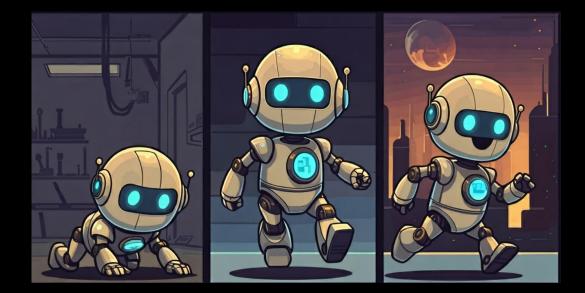
Combine deterministic approaches with new probabilistic capabilities



Crawl, walk, run

Find the sweet spot of what's possible now and will be possible soon

But be ready for what's coming



Demand for good software is

near<u>ais</u>∞

- Reduce toil for migrations
- Increase test coverage
- Give humans more options
- Help humans think about problems



<u>Humans</u>

- More time building
- Focus time on the craft of software engineering
- Break down complex problems
- Define architecture
- Set best practices

Questions?



Adam Huda Sr. Eng Manager Al Foundations & Developer Experience





Ty Smith Principal Eng Developer Platform

