



Improving Software Quality Using Machine Learning and AI

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Software Development Challenges

- Increased complexity
- Heavy reliance on people
- Lack of automated tools
- Time to market pressure
- Emerging technologies
- QA trade-offs



Software Maintenance

70% of the overall development cost

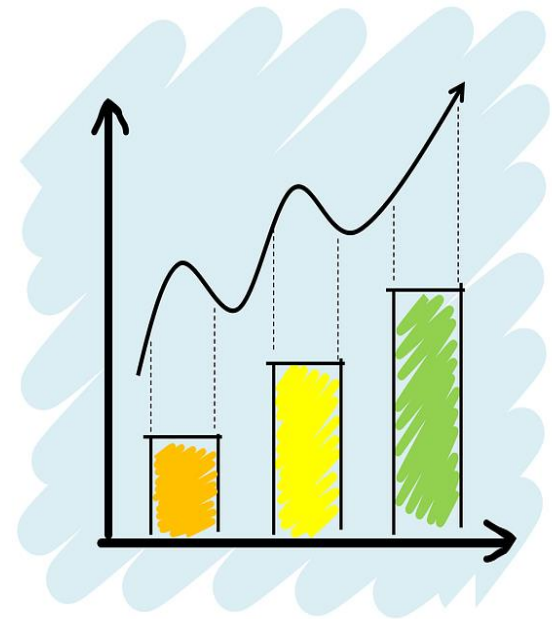
Up to 50% of maintenance cost is on fixing bugs

Bugs may have **serious consequences**

Defects cost the economy **billions of \$** annually

Emergence of Software Analytics

- Data-driven SW development and maintenance
- Big Data: source code, bug reports, test cases, logs, user feedback, etc.
- Predictive analytics using ML, DL, CI, and PR
- Information visualization of large-scale data

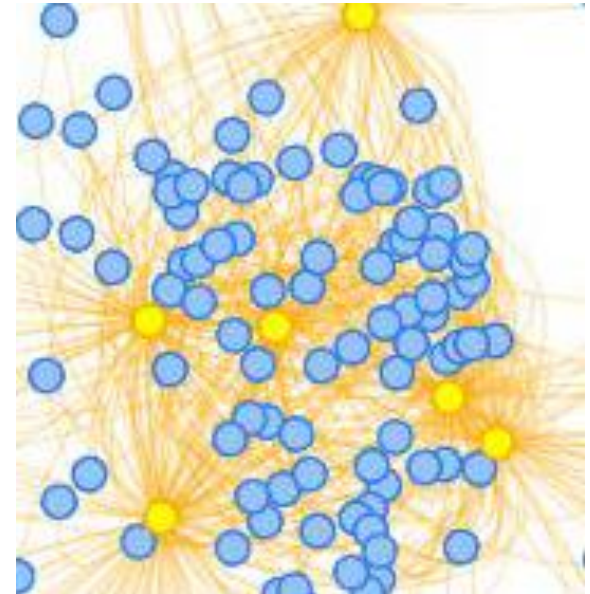


Defect Detection/Prediction Research

- Defect Prediction
 - Statistical analysis
 - Call-graph analysis
 - Analysis of code changes
 - Leverage of historical data
- Automated Patch Generation
 - Development of fixing patterns
 - Reuse of human written patches
 - Directed patches towards specific bug types

Problems with existing techniques

- Offline processing (after the code is built)
- Presence of the entire source code
- Extensive setup and high learning curve
- Lack of clear actions to developers
- High rate of false positives



Our solution: CommitAssistant

- A prototype tool resulting from an NSERC research project between my research lab at Concordia University and Ubisoft Laforge
- Main Features:
 - Detection of bugs at commit-time, i.e., as programmers write code
 - Supports multiple programming languages
 - No external tools or setup required
 - Leverage of historical bugs and fixes
 - High TRL

CommitAssistant Phases

1

Train models of historical defect and healthy commits and associated code

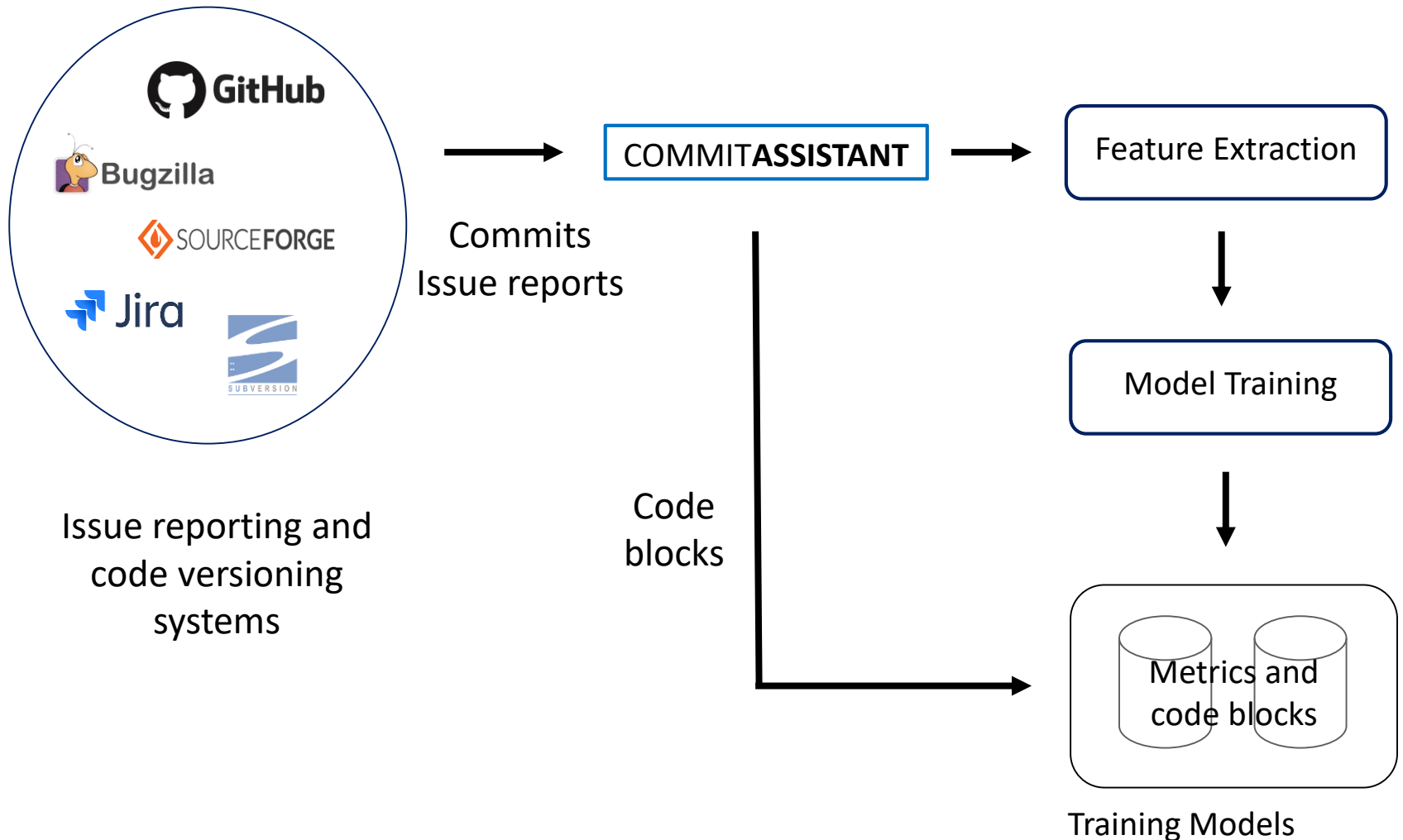
Intercept and analyze developers' commits before they reach the central code repository

2

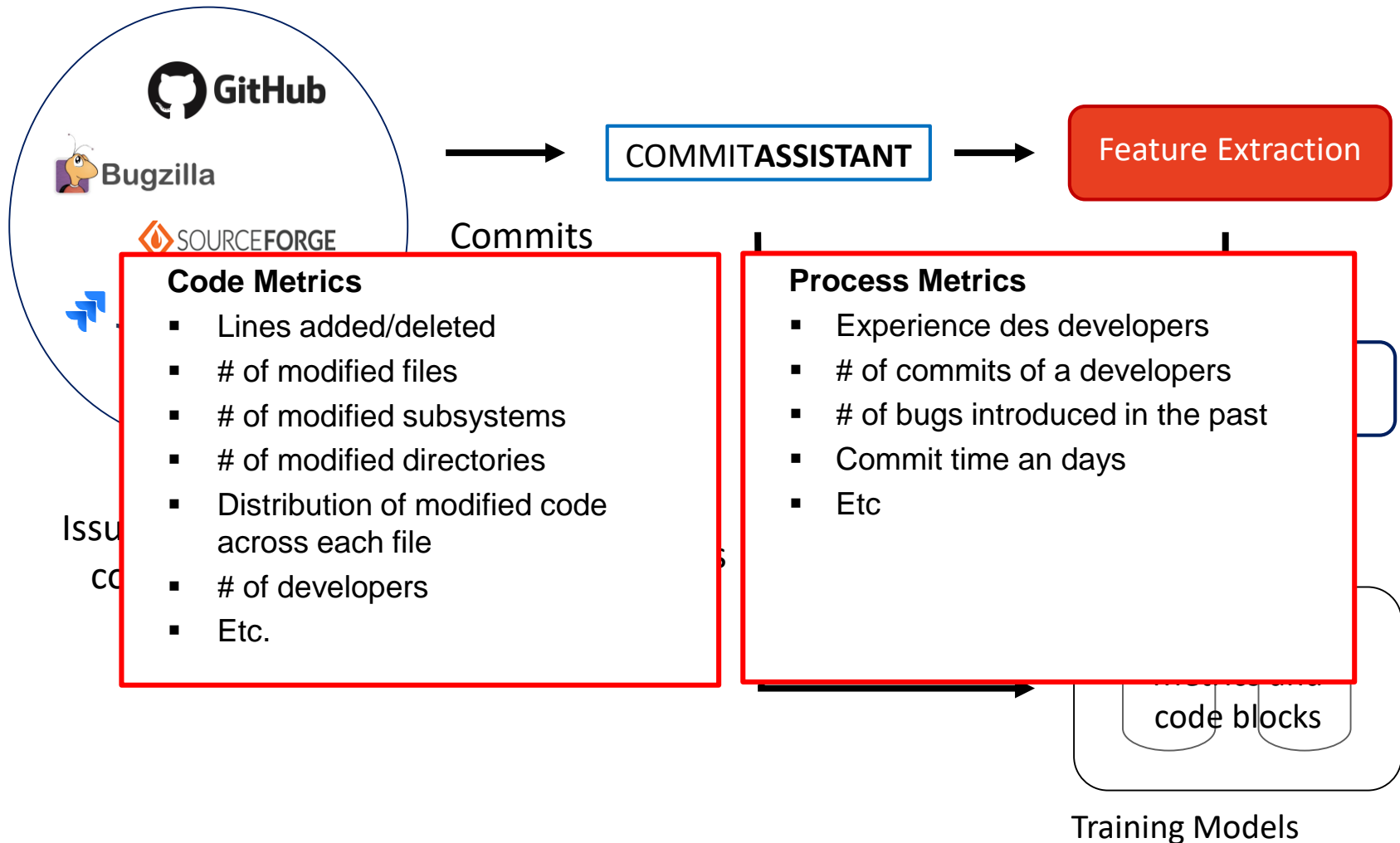
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Notify developers and **propose fixes** for risky commits

Step 1: Train models



Step 1: Train models



Steps 2, 3: Analyze commits and propose fixes

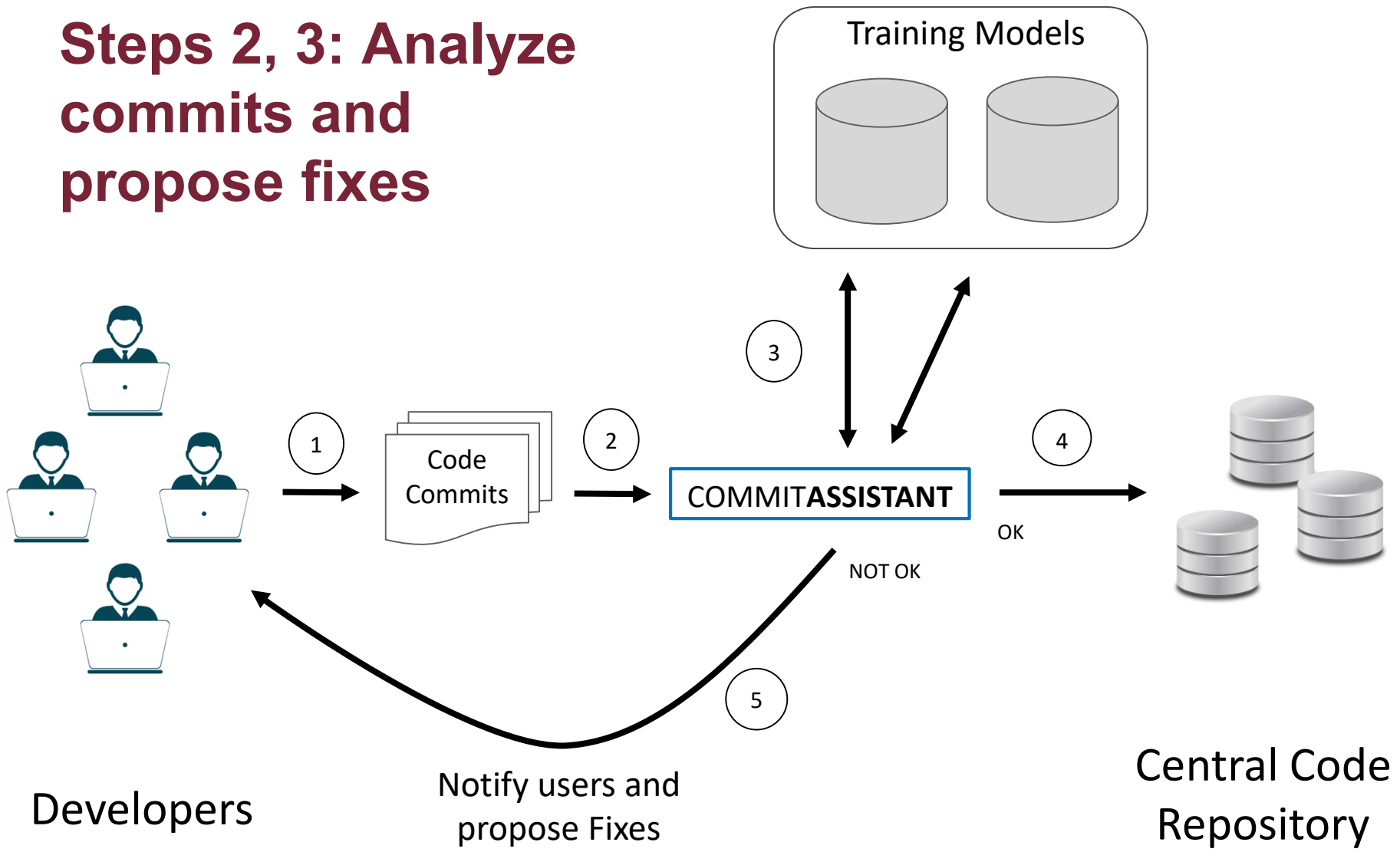


TABLE 3: BIANCA results in terms of organization, project name, a short description, number of class, number of commits, number of defect introducing commits, number of risky commit detected, precision (%), recall (%), F₁-measure (%), the average similarity of first 3 and 5 proposed fixes with the actual fix and the average time difference between detected and original.

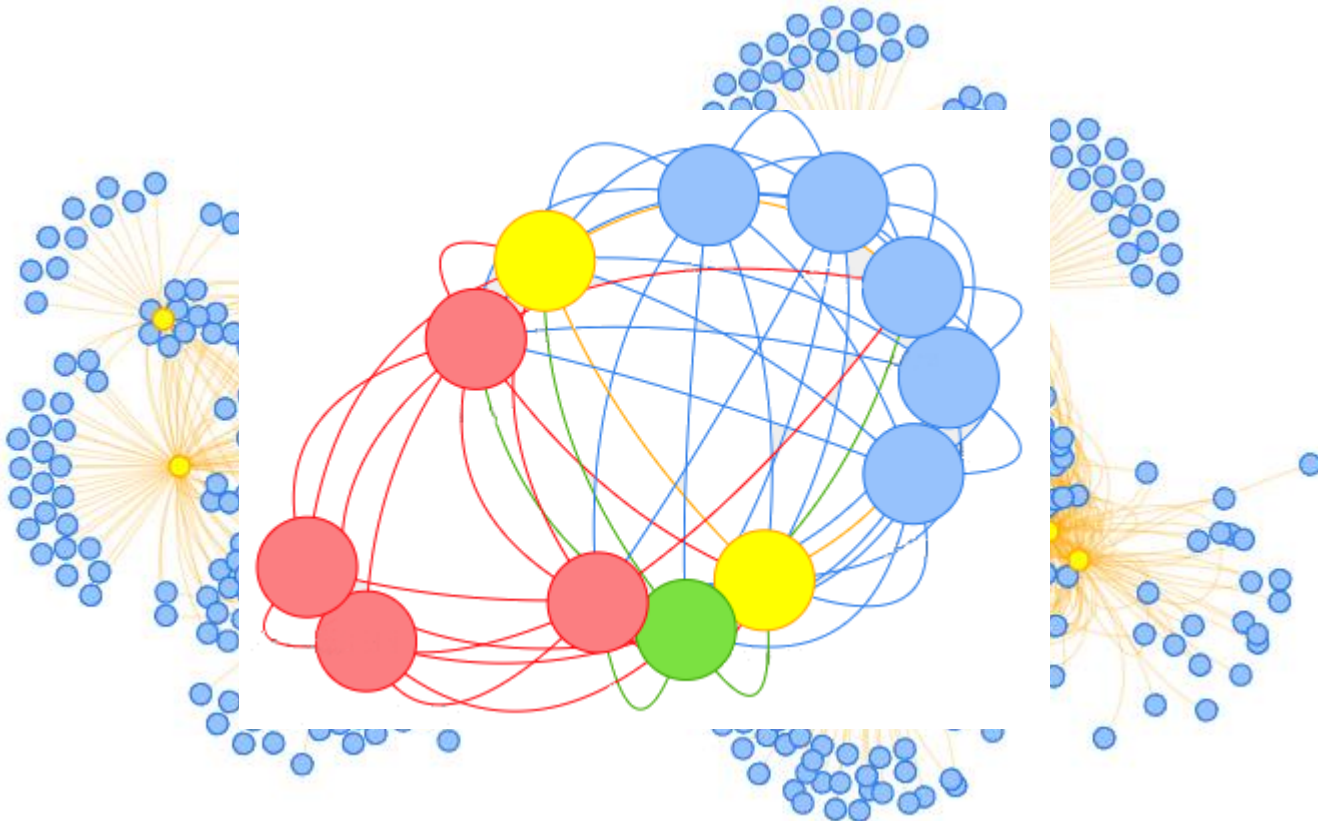
Organization	Project Name	Short Description	NoC	#Commits	Bug Introducing Commit	Detected	Precision	Recall	F ₁	Top 5 Fixes Similarity	Top 3 Fixes Similarity
Alibaba	druid	Database connection pool	3,309	4,775	1,260	787	88.44	62.46	73.21	39.97	46.69
	dubbo	RPC framework	1,715	1,836	119	61	96.72	51.26	67.01	60.01	57.14
	fastjson	JSON parser/generator	2,002	1,749	516	373	95.71	72.29	82.37	18.19	15.23
Apache											48
Cloj											58
Droy											10
Eclig											52
Exci											56
Face											82
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Square	okhttp	HTTP+HTTP/2 client	344	2,649	592	474	93.04	80.07	86.07	29.09	24.91
	okio	I/O API for Java	90	433	40	24	100.00	60.00	75.00	31.51	35.50
	otto	Guava-based event bus	84	201	15	15	93.33	100.00	96.55	54.11	49.94
	retrofit	Type-safe HTTP client	202	1,349	151	111	99.10	73.51	84.41	49.88	45.46
StephaneNicolas	robospice	Android library	461	865	113	39	87.18	34.51	49.45	60.90	65.04
ThinkAurelius	titan	Graph Database	2,015	4,434	1,634	527	90.13	32.25	47.51	48.64	50.59
Xetorthio	jedis	Redis client	203	1,370	295	226	92.04	76.61	83.62	25.69	29.45
Yahoo	antheion	Plugin for Apache Nutch	1,620	7	0	-	-	-	-	-	-
Zxing	zxing	1D/2D barcode image	3,030	3,253	791	123	94.31	15.55	26.70	29.35	37.96
Total			96,003	165,912	41,225	15316	90.75	37.15	52.72	40.78	44.17

Evaluation

- 42 open source projects
- Precision = 90%
- Recall: 37%
- 79% of the proposed fixes are accurate

Project Clustering

We can improve the detection accuracy if we search within inter-related projects



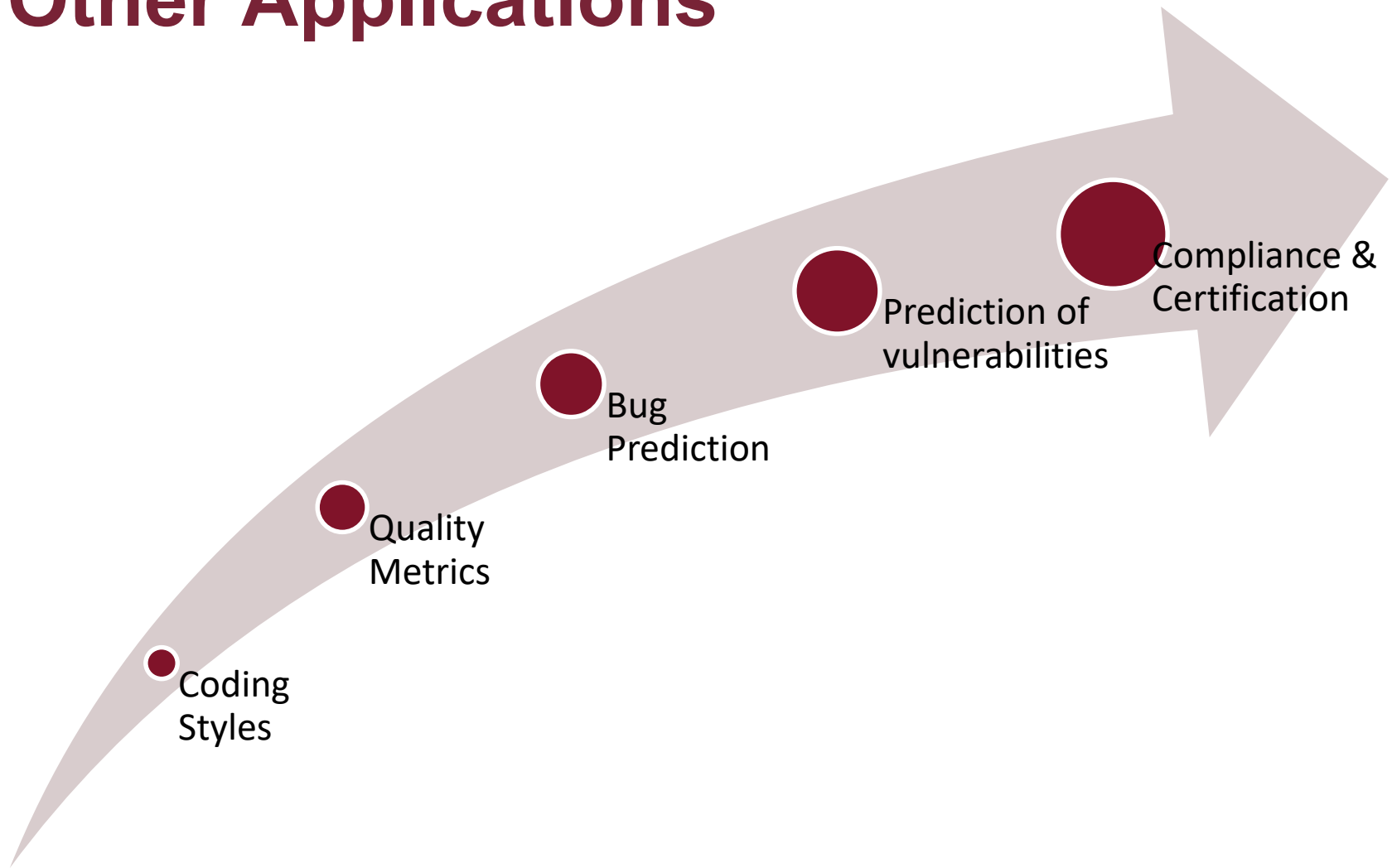
Evaluation of CommitAssistant at Ubisoft

- 12 Ubisoft AAA games
- 10+ millions of LOCs
- Precision = 79%
- Recall = 65%
- 67% of the fixes were deemed acceptable

Impact

- Commit-Assistant is designed **to integrate** well with the workflow of Ubisoft developers
- Clever-Commit (production version of Commit-Assistant) **is widely deployed** at Ubisoft
- Ubisoft announced in a press release that Commit-Assistant can **cut the bug fixing time by 20%**
- **Mozilla** announced that it is working with Ubisoft to use Clever-Commit in the **development of Firefox**

Other Applications



CommitAssistant as JIT Monitoring Tool

Analyzing commits provides real-time view of code quality:

- # of introduced bugs
- File metrics
- Subsystem metrics
- Code change density
- Code complexity
- Number of fixes
- Etc.



Open Questions

- How to reuse this technology in other areas such as avionics and aerospace?
- How can we apply CommitAssistant to embedded and critical safety systems?
- What is the interplay between commit analysis, testing, operational intelligence, etc.?
- Can this technology help with certification and compliance of software?
- Is this technology certifiable?

Conclusion

- Machine learning and AI are needed to reduce overhead of bug fixing
- CommitAssistant:
 - reuses existing knowledge and AI to improve new code
 - improves quality by providing early feedback to developers
 - assists developers on how to fix risky commits
 - works well on Ubisoft systems



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