

1 FSE 2026 Workshop: Data Intensive Software Development
2 (DISD)
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7 **ACM Reference Format:**

8 Anonymous Author(s). 2025. FSE 2026 Workshop: Data Intensive Software Development (DISD). 1, 1 (October
9 2025), 6 pages. <https://doi.org/10.1145/nnnnnnnn.nnnnnnn>

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11 **1 Introduction**
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13 Modern software systems are becoming increasingly data-intensive. They process massive quantities
14 of data and also produce large-scale, machine-generated data, such as traces and telemetry, which
15 is used for runtime monitoring, log analysis, debugging, and testing. The analytical pipelines are
16 also performance-critical, and their behavior is highly dependent on the scale and content of the
17 data they process.

18 Due to this massive scale, it is essential to incorporate intelligent AI agents that can not only
19 process the data but also reason about it within the context of the software. Integrating these
20 AI-driven, data-informed feedback loops is the key to enabling autonomous software operations
21 and providing context-aware guidance to developers. However, achieving trustworthy AI-native
22 software engineering requires addressing critical challenges that directly impact the robustness,
23 reliability, and trustworthiness of data-intensive systems

- 24 • Data-dependent uncertainty: The dynamic nature of large-scale, machine-generated data
25 makes it difficult to design comprehensive test cases and debug unexpected behaviors.
26 Further, the use of AI and agentic approaches, both (1) in these software systems and (2)
27 for debugging and monitoring large-scale data for detecting anomalies, can amplify the
28 uncertainty in diagnosis.
- 29 • Massive scale: The massive volume of data overwhelms traditional methods, making analysis
30 computationally infeasible. This requires systems-level optimization for storage, tracing,
31 and monitoring as critical system behaviors only manifest at production scale.
- 32 • Stringent latency requirement: The real-time or near real-time requirements of data-intensive
33 applications make traditional offline analysis too slow to detect and respond to issues timely.
34 The long running time of data-intensive applications makes techniques such as fuzzing not
35 easily applicable at scale.
- 36 • Robustness under evolving conditions: Data-intensive systems operate in dynamic envi-
37 ronments where both data distributions and infrastructure conditions shift unpredictably.
38 Trustworthy AI approaches must remain stable and reliable despite distribution drift, adver-
39 sarial inputs, and noisy runtime data.

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48 ACM XXXX-XXXX/2025/10-ART

49 <https://doi.org/10.1145/nnnnnnnn.nnnnnnn>

50 2 Workshop Information

51 This workshop community will provide a forum for researchers and practitioners to advance soft-
 52 ware engineering research around data-intensive software development. We will invite discussions
 53 that explore fundamental ideas, practical solutions, and cross-domain innovations in this space,
 54 including but not limited to:

- 55 • Data-intensive software testing, debugging, runtime monitoring, and log analytics
- 56 • Semantics lifting from systems-generated data
- 57 • Modelling application behavior via data-system coordination
- 58 • The use of AI, ML, and agentic approaches for monitoring, debugging, and testing systems-
 59 generated data
- 60 • Algorithms and foundations for testing and debugging performance-critical systems
- 61 • Benchmarks for testing, debugging, and analysis of systems-generated data

63 *Goal:* These research topics are vital for trustworthy AI incorporation in data-intensive software
 64 development. Due to data sensitivity, agent systems' interaction logs, large-scale logs, input data,
 65 queries, workloads for big data analytics, and associated bugs, are often not released to the public. To
 66 drive **trustworthy AI and agentic AI incorporation in data intensive software development**,
 67 we are in the urgent need of forming a research community (a consortium) that **curates and**
 68 **maintains industry-scale, synthetic datasets and benchmarks**.

69 *Organizers.*

- 71 • Lionel Briand (lbriand@uottawa.ca), Lero Centre, Ireland
- 72 • Tse-Hsun (Peter) Chen (peterc@enccs.concordia.ca), Concordia University, Canada
- 73 • Muhammad Ali Gulzar, (gulzar@cs.vt.edu), Virginia Tech
- 74 • Yintong Huo (ythuo@smu.edu.sg), Singapore Management University
- 75 • Miryung Kim (miryung@cs.ucla.edu), UCLA
- 76 • Michael Lyu (lyu@cse.cuhk.edu.hk), Chinese University of Hong Kong
- 77 • Weiyi Shang (wshang@uwaterloo.ca), U. Waterloo

79 *Format and required services.* The proposed workshop will be structured as a 1-day event de-
 80 signed to maximize participation, knowledge exchange, and collaborative problem-solving around
 81 data-intensive software. The program will feature paper presentations in the nature of research
 82 idea proposals or position papers), along with interactive group activities designed to stimulate
 83 discussion and idea generation.

84 **Keynote session (morning 1).** The morning session of the workshop will begin with a keynote,
 85 setting the stage by outlining the major challenges and opportunities in data-intensive software
 86 engineering. This shared foundation will help align participants around the central themes of the
 87 workshop while sparking ideas for further discussion. We plan to invite keynote speakers who can
 88 discuss the scale of AI incorporation in data intensive software development in industry such as
 89 Tim Kraska (Amazon Web Services), Dongmei Zhang (Microsoft), Nachiappan Nagappan (Meta),
 90 Chao Peng (ByteDance), Ahmed Hassan (Huawei), Jeromy Carriere (DataDog).

91 **Paper presentation session (morning 2).** Following the keynote, in the second session in the
 92 morning participants will present their papers. We will be soliciting papers in the type of research
 93 idea proposals or position papers to stimulate discussions. These presentations will provide concrete
 94 starting points for the collaborative activities scheduled later in the day.

95 By the end of the morning session, the organizers will finalize the themes that emerged from
 96 the keynote and proposal presentations. The afternoon breakout groups will be structured around
 97 well-defined themes of interest.

99 **Breakout session (afternoon 1).** Based on the consolidated four to five themes created by
100 the end of the morning session, and attendees will divide into smaller breakout rooms. Within
101 each group, participants will explore shared challenges, identify synergies, and generate new ideas
102 that may not surface in larger plenary discussions. At the end of this breakout session, our goal is
103 to form a team to volunteer to write a section in a community driven article on **challenges and**
104 **reflections in data-intensive software development.**

105 **Round-table session (afternoon 2).** After the breakout discussions, all groups will reconvene
106 in the main session to share key insights and perspectives across themes. Each group will briefly
107 present their key findings and emerging ideas. The round table format will encourage an open,
108 interactive exchange, highlight connections between themes, and debate differing viewpoints. The
109 expected deliverable for this round-table session is a **cohesive outline for a community-driven**
110 **article** on challenges and reflections in data-intensive SW development.

111 **Publicity Plans.** Workshop publicity will span multiple coordinated channels. Announcements
112 will be distributed via SIGSOFT's SEWORLD mailing list to reach the broader software engineering
113 community. We will regularly post updates on X and LinkedIn to highlight key dates for the
114 workshops and feature our keynote speakers. Additionally, we will promote the workshops through
115 FSE's official channels, including the conference website and social media platforms. We will also
116 consider sending direct invitations to FSE 2026 authors across various tracks to encourage their
117 participation, submissions, and attendance at the workshop.

118 **Publications** The workshop organizers will not be submitting papers to their own workshop.
119 After the workshop, the organizers will summarize advances and challenges in the area of data-
120 intensive software development as a recurring **community-driven article** that forms the body of
121 knowledge in this area. We have received a positive response from the editor of chief at IEEE Soft-
122 ware on disseminating the workshop outcome report, as a recurring featured article on "Challenges
123 and reflections on data intensive SW development."

124 **Statement of Overlap With Prior Workshops** This workshop is different from a continuing
125 20 years history of Mining Software Repositories¹ community by focusing on systems-generated
126 data. The AIOps workshop² focuses on applying AI to cloud operations, while our workshop is not
127 limited to cloud operations but building, debugging, and testing data-intensive SW development
128 and AI incorporation in such context. Also our goal is to create a community to curate and
129 maintain industry-scale synthetic dataset and benchmarks to drive research in AI incorporated,
130 data-intensive SW development. Unlike the Workshop on Distributed Software Development,
131 Software Ecosystems and Systems-of-Systems³, which emphasizes distributed development and
132 ecosystems, this workshop target the complexities of large-scale data processing. Compared to the
133 International Workshop on Agentic Engineering⁴, which is centered on autonomous agents, our
134 scope is broader and data-driven. While LLM4Code⁵ is focused on large language models for code-
135 centric tasks, this workshop address challenges in building, debugging and testing data-intensive
136 pipelines.

137 3 Paper Selection Procedure

138 The workshop welcomes the following submission types:

- 140 • Position Statements (1 to 2 pages): early-stage ideas, industrial perspectives, proposals to
141 contribute to the working group report.

142 ¹<https://www.msrgconf.org>

143 ²<https://cloudintelligenceworkshop.org>

144 ³<https://conf.researchr.org/home/sesos-wdes-2021>

145 ⁴<https://conf.researchr.org/home/icse-2026/agent-2026>

146 ⁵<https://llm4code.github.io>

- 148 • Short Papers (up to 5 pages): early-stage ideas, visions, or tool reports.
- 149 • Full Papers (6–10 pages): novel approaches, frameworks, or evaluations.
- 150 • arXiv Presentations (non-archival): recent preprints for open discussion.

151 Authors of accepted position statements will collaborate to combine and expand their experience
 152 and insight into a community paper to be submitted to IEEE Software or a similar venue. The goal
 153 of this paper is to provide a shared perspective on the current state of the field and offer a roadmap
 154 for the future.

155 All papers will be submitted via HotCRP and will be reviewed in a double-blind manner. The
 156 submission should comply with the ACM format (in line with FSE 2026) and should present the
 157 original contribution. At least one author of each accepted paper must register for the workshop
 158 and present the paper there.

159 All accepted papers, except for the position statements and arXiv presentation, will appear in the
 160 FSE 2026 workshop proceedings by default. For the non-archival papers and position statements,
 161 the camera-ready version will only be posted/advertised on our workshop website. Please note
 162 that no matter which option you choose (archival or non-archival), the submission should be fully
 163 original (not accepted/published anywhere else) by the submission time, and at least one author has
 164 to register for the workshop and present. The official publication date of the workshop proceedings
 165 is the date the proceedings are made available by ACM.

166 **TODO: Expected attendees**

167 **TODO: Format: 1 day?**

168 **TODO: PC formation and Publicity: Ian + Yintong + Gulzar, merge with keynote list**

169 **TODO: merge the two lists** Domenico Bianculli (University of Luxembourg),

170 Michael Pradel (CISPA),

171 David Lo (Singapore Management University),

172 Xiaofei Xie (Singapore Management University),

173 Lingming Zhang (UIUC),

174 Chao Peng (Bytedance),

175 Ying Li (Peking University),

176 Jie Zhang (KCL),

177 Pinjia He (Chinese University of Hong Kong, Shenzhen)

178 Shan Lu (University of Chicago)

179 Heng Li (Polytechnique Montreal)

180 Junwen Yang (Meta)

181 Jeromy Carriere (<https://www.linkedin.com/in/jeromycarriere/>)

182 **TODO: miryung:we probably need some more industry participants? more european participants?**

184 **4 Working Group Titles for "What's going on in data intensive SW development"** 185 **curation paper**

187 **4.1 Working Group 1: Data-Intensive Benchmarks, Bugs, and Oracles**

188 Data-intensive software systems disproportionately lack benchmarks, bug repositories, and domain-
 189 specific oracles. For example, widely used frameworks such as Apache Hadoop and Spark are still
 190 evaluated against decades-old SQL benchmarks (e.g., TPC) or unrealistically simple programs
 191 (e.g., sort). At the same time, we have little systematic knowledge of the bugs that occur in these
 192 frameworks and the applications built atop them, nor of the oracles typically used to detect such
 193 bugs. As a result, software engineering research in this domain has remained underrepresented. A
 194 core objective of this workshop is to stimulate open discussion on pathways to design natural, real-
 195 world-inspired, and scalable benchmark programs for data-intensive stacks, their bugs, and their

197 oracles, with the aim of advancing *correctness, performance, and scalability* testing. A key element
198 for this working group is to investigate synergistic approaches to acquire such benchmarks from
199 commercial and industry stakeholders (e.g., Amazon EMR, Google Dataflow, Databricks, Snowflake)
200 while respecting intellectual property, data privacy, and business confidentiality constraints.

201 TODO: homework: title and one sentence description

202 TODO: Data-intensive Benchmarks Gulzar Miryung, bugs oracles,

203 TODO: Synergy with AI/ Agentic Peter

204 TODO: AI-Ops, runtime monitoring, Yintong

206 **4.2 Working Group2: Data-Intensive Software Runtime Monitoring**

207 Modern software systems generate vast amounts of runtime data, but monitoring often treats sys-
208 tems as a monolith, obscuring the component interactions necessary for precise failure attribution.
209 This challenge is further amplified by the integration of recent black-box AI/LLM plugins, which
210 complicates root cause analysis. This working group will tackle these challenges by exploring meth-
211 ods to *instrument and exploit runtime data for monitoring*. We will focus on creating benchmarks
212 and analyzing techniques for fine-grained failure attribution and localization, especially within
213 complex systems that combine traditional software with AI components.

215 **4.3 Working Group X: data-intensive systems debugging and testing**

216 Debugging and testing data-intensive systems present unique challenges that set them apart from
217 traditional software. The scale, heterogeneity, and continuous flow of data make it difficult to
218 reproduce failures, isolate root causes, or ensure comprehensive test coverage. Data quality itself
219 can become a hidden source of errors, where issues may arise not from the code but from anom-
220 alies, inconsistencies, or biases in massive datasets. Moreover, the interplay between distributed
221 components and complex data pipelines creates emergent behaviors that are hard to anticipate
222 with conventional testing techniques. TODO: I wonder wehther we should combine 4.1 and 4.3

224 **4.4 Working Group X: Non-functional quality attributes of data-intensive systems**

225 Non-functional aspects of data-intensive systems, such as security, privacy, performance, scalability,
226 reliability, and energy efficiency, are central concerns, as they directly shape the trustworthiness
227 and usability of these systems. Managing massive and heterogeneous data at scale requires not
228 only functional correctness but also careful attention to how systems safeguard sensitive infor-
229 mation, maintain resilience under heavy loads, and deliver predictable performance in dynamic
230 environments.

232 **4.5 Working Group X: AI-Agentic Systems as Data Producers and Consumers**

233 TODO: to be changedThe rise of AI-agentic systems creates a unique dual role: they are both sophis-
234 ticated consumers of data-intensive infrastructure and significant generators of complex behavioral
235 data. This working group will explore the complete lifecycle of data in AI-agentic environments,
236 where the interactions, decisions, and reasoning traces of AI agents themselves become valuable
237 telemetry for system understanding. We will address challenges such as instrumenting multi-agent
238 workflows, capturing meaningful agent decision logs, and leveraging this AI-generated data to
239 improve system reliability, performance, and transparency. Key focus areas include developing
240 standards for agent telemetry, creating benchmarks that include agent interaction patterns, and
241 designing monitoring solutions that can attribute system behavior to specific agent decisions
242 or emergent collective behaviors. This group will bridge the gap between traditional software
243 monitoring and the new paradigms required for autonomous, AI-driven systems. TODO: keep the
244 title only and move the merged content or content on the website TODO: Gulzar:website TODO:
245

246 Ian: differentiation from existing work TODO: Yintong, Peter, Me taking a pass in the oder, will
247 have a final version before Oct 3 meeting

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