
Analyze Conference Venue

Yiqiao Yin
W.Y.N. Associates, LLC

Abstract

1 This assignment investigates and reviews the top software engineering
2 conferences. These conferences include ICSE, MSR, FSE, ASE, and
3 ISSTA. For each of the listed conferences, the report reviews the main
4 contributions and components of that conference.

5 **1 ICSE**

6 **1.1 Brief**

7 The International Conference on Software Engineering or ICSE is an annual international
8 conference focusing on advanced research in software engineering. The first ever ICSE
9 meeting was scheduled in Washington DC, USA and it was sponsored by the National
10 Bureau of Standards and IEEE Computer Society. The ICSE also has an A rating from
11 the Rankings of the Computing Research and Education Association of Australasia.

12 **1.2 Tracks**

13 The ICSE has the following tracks: Technical Track, Software Engineering in Practice
14 of EIP track, Joint Track on Software Engineering Education and Training or JSEET,
15 software Engineering in Society or SEIS, New Ideas and emerging Results or NIER,
16 Demonstrations or DEMO, ACM Student Research Competition or SRC, Artifact Evaluation
17 or AE, Doctoral Symposium or DS, New Faculty Symposium or NFS. Each track
18 has keynote speakers, posters or papers, presentations, and awards. This is very well
19 funded and thoroughly organized on an international level. It would be a great honor if
20 one gets their paper or work selected by ICSE.

21 **1.3 Selected Work: Gopinath et al. (2014)**

22 An important field in software engineering is the code testing. The work of Gopinath
23 et al. (2014) discussed the existing problem of code test and search of the potential faults
24 in a software program. In a perfect world, it is desirable to have the measure set up to
25 detect the exact the faulty component in a code structure. The authors' work has two
26 major motivations. The results needs to be applicable on large-scale set of real-world
27 software programs and the programs in the database need to be diverse enough to be
28 able to allow sufficient inference and generalization. The second motivation is to achieve
29 statistically significant result. The proposed mutation testing delivers high performance.
30 A long term goal that increase the desirability of mutation test is to develop alternative
31 methods to detect software faulty components with the state-of-the-art accuracy. The

32 author's vision is quite forward. However, it is extremely challenging to keep this
33 promise.

34 **1.4 Selected Work: Inozemtseva and Holmes (2014)**

35 The above research proposed mutation test to serve as a test suite to detect faulty
36 programs. In the same field, another team presents novel research of the relationship
37 between test suite size, coverage and effective Inozemtseva and Holmes (2014). Their
38 work seeks the answers to understand whether the test suite is correlated with a test case
39 in the suite. In addition, their work also explores the ground of the effective of a test
40 suite with regards of the statement coverage. In their work, the first contribution is a
41 comprehensive survey of the investigation focusing on the correlation or association
42 between coverage and effectiveness. According to their study, there is evidence show that
43 there exists a low to moderate level of correlation between coverage and effectiveness
44 assuming that the suite size is controlled. The value of their work is to initiate future
45 research that can confirm the findings in real faulty programs that poses threat to validity.

46 **1.5 Selected Work: denaro2015dynamic**

47 Another interesting field is text generation. In this field, research in data flow testing
48 has drawn a lot of attention in recent years. The classical data flow test which is static,
49 however, fails to assist the software programmers to identify the relations, because
50 the code flow is more dynamic than before. Hence, the work by Denaro et al. (2015)
51 proposed a novel method called DynaFlow, which is new approach to generate test cases
52 that takes the advantage of the useful data flow information where this was originally
53 difficult to compute. To work with not just general programs but also interprocedural
54 testing where some programs are object oriented programs, the design of DynaFlow is
55 proposed to tackle dynamically allocated data flow.

56 **2 MSR**

57 **2.1 Brief**

58 The Mining Software Repositories or MSR holds its significant role in the field of data
59 science, machine learning, and artificial intelligence in software engineering. The pur-
60 pose and the motivation for MSR is to improve the development of software engineering
61 practices using vast amount of data. These practices can involve source control systems,
62 defect tracking systems, code review repositories, and so on. The conference encourages
63 novel ideas in software development utilizing data science and machine learning, which
64 is quite important at today's economy.

65 **2.2 Tracks**

66 MSR has many representative tracks. These tracks include technical work and research,
67 mining challenges, and data showcase. In addition, hackathon is hosted on an annual
68 basis to improve the development of the cybersecurity. The conference, just like ICSE,
69 also offers keynote speakers as well as awards for each track session to encourage and
70 incentivize scholars to make distinct contributions in the field.

71 **2.3 Selected Work: Malhotra et al. (2021)**

72 The work by Malhotra et al. (2021) investigate and compare the vast amount of feature
73 reduction techniques in software change prediction. The field of software change
74 prediction or SCP aims to explore the various structural metrics inside a software
75 program and try to determine whether the future generation of the software is subject to
76 changes and updates. The forecast or prediction conducted in the field largely depends on
77 the correct specification of features which requires the researchers deep understanding
78 and comparative analysis of different feature reduction or FR techniques to address
79 the problems under high dimensionality, feature irrelevance, and feature redundancy
80 problems.

81 **2.4 Selected Work: Ali et al. (2020)**

82 This paper is presented in the 17th MSR with a fancy title “Cheating Death” Ali et al.
83 (2020). The authors developed a survival model to analyze the usage inactivity over a
84 time span of 165 months on more than 3,000 python projects. It is the most interesting
85 projects in this report and it is not surprising that the paper made it in the 17th MSR.
86 The work has an extremely attractive motion. Computer scientists, data scientists, and
87 computer engineers all rely on open source software or OSS which are public projects
88 available for use. For machine learning and deep representation learning, it is common
89 to use pre-trained models as a start. These projects tend to have longer survival time.
90 However, for projects that are difficult to elaborate or to solve, it is challenging for
91 developers to come back to the project and renovate it. For these projects, they are
92 expected to have shorter survival time. The work done by this team shed lights to the
93 effects of attributes of open source project and the overall length of activity level that an
94 open source project could have.

95 **2.5 Selected Work: Tufano et al. (2018)**

96 Like the similarity score between two articles, the source code analysis provides sim-
97 ilarity score between each component of code blocks an essential understanding of
98 the software engineering tasks. This technique is important at clone detection, impact
99 analysis, refactoring and so on. The similarity score or code similarity is a quantitative
100 measure between two blocks of code from one or different software programs that con-
101 sists of features automatically detected machines or hand-craft features by programmers.
102 Recent development of deep learning or DL models can construct effect machines to
103 replacement some of the cumbersome human labors. The work by the author Tufano
104 et al. (2018) show state-of-the-art performance of a deep learning pipeline that detects
105 and forecasts the similarity score.

106 **3 FSE**

107 **3.1 Brief**

108 Foundations of Software Engineering or FSE is a world renowned platform for re-
109 searchers, scholars, practitioners, and educators to present and showcase their latest
110 innovations and research practices. The conference consists of many different field and
111 invite different scholars together every year.

112 **3.2 Tracks**

113 The conference starts with the plenary events and these events will have keynote speakers
114 invited who are top of their fields. The conference, unlike the previously discussed
115 ones, has a diversity and inclusion events to emphasize women in science and minority
116 research groups which is quite important at today's world. The main events kick off
117 with the main research track and industry track. The research track consists of work
118 from academic or university labs while the industry track consists of work from labs in
119 the industry. The conference also has a designated doctoral symposium for graduate
120 students to showcase their ongoing research. The conference has a student research
121 competition and workshop every year for junior level research.

122 **3.3 Selected Work: Nguyen et al. (2014)**

123 This work Nguyen et al. (2014) explores the dependence relations between application
124 programming interfaces or APIs from online open source libraries. The central idea of
125 the authors' work is to understand the preconditions of the APIs that would be used for
126 ultra-large scale code structures with pre-defined built-in modules. To understand this
127 task, the first step is to search for all client methods are used for invoking the APIs. Then
128 it is essential to compute the potential conditions used in order to achieve the calls of
129 certain libraries. Their work interpreted 120 million lines of code from the standard Java
130 Development Kit or JDK library and delivered results with high recall and precision,
131 79% and 84% respectively. The paper sets important milestones of the idea of mining
132 API references which is very important to today's coding standard considering that APIs
133 are used everywhere in most of the components of today's world such as data science,
134 machine learning, and deep learning. Based on the analysis, their work expects that
135 the APIs to have preconditions that can make appearance in the large-scale corpus of
136 open source libraries very frequently. Moreover, sometimes it is possible to predict the
137 preconditions as well as the dependence relationship amongst different APIs before the
138 library are called, which is very useful in software engineering.

139 **3.4 Selected Work: Shi et al. (2014)**

140 In large test suites, the regression testing is a fundamental tool that benefits most of the
141 software develop environment. However, conventional regression testing suites share
142 three difficult properties that pose challenges to move above and beyond. The first
143 property is that the requirements of the test suites are governed by pre-defined coverage
144 suites. The second property is that for scenarios where developers use reduced test suites
145 the environment has to be tailored in order to satisfy all former requirements. The last
146 property states that the quality of the software testing suites are described and written
147 according to the reduced test environment. These properties posed great challenges to
148 develop efficient software engineering environment for regression testing suites. The
149 authors develop and explore the reduced test suites aiming to develop reduced test suites
150 with properties such that they enable efficient development environment without having
151 to be limited to the three properties.

152 **3.5 Selected Work: Huo and Clause (2014)**

153 This paper solves several issues in writing oracle machines. Oracle machines are another
154 form of the Turing machine that is connected with an oracle. The oracle, in this case
155 (not to be confused with the Oracle Script in Ancient China), is an entity of a black-box
156 algorithm that is capable of solving certain problems. There are two major types of

157 problems: a decision problem and a function problem. A decision problem is the type of
158 problems that is written as a set of natural numbers or a set of strings. Their solutions
159 usually can be answered with “yes” or “no”. A function problem is written by a function
160 from a natural set of numbers. The solutions to function problems are usually value or
161 some numerical output of a function of the set of numbers defined in the oracle. As
162 a consequence, developers who are responsible for writing oracles or similar entities
163 cannot always thoroughly check each step of the machine because the entity acts like a
164 black-box in the pipeline. The testing procedures are fragile and can be, sometimes,
165 full of bugs. The debugging procedure is not that easy to handle either due to its dynamic
166 structure or complicated developmental pipeline. The author proposed a technique that
167 address these issues using two brittle inputs. The first type of brittle inputs are the type of
168 assertions that has dependent relationship with uncontrolled inputs. The second type of
169 the brittle inputs are the unused inputs which are the type of inputs that are not checked.
170 These two types of inputs allow the users and the developers to automatically analyze
171 test oracles.

172 **4 ASE**

173 **4.1 Brief**

174 The international conference Automated Software Engineering or ASE is a world-
175 class research platform that is focusing on software engineering. The conference is
176 held annually at a different location every year and it brings together the leaders and
177 researchers of the world. The conference targets on topics such as analysis, design,
178 implementation, testing, and software maintenance and so on.

179 **4.2 Tracks**

180 The conference has the following tracks. The main track, as a highlight, is the artifact
181 evaluation, which focuses on evaluation of the most recent research in the selected topics
182 such as software design, software implementation, and testing procedures and so on.
183 Like the conferences above such as ICSE, MSR, or FSE), this conference also has a
184 doctoral symposium to show case doctoral research and an industry showcase which
185 provides opportunity for industry practitioners. Unlike the previous conferences, this
186 conference has a specific gaming session called the ASE4Games.

187 **4.3 Selected Work: Tantithamthavorn and Jiarpakdee (2021)**

188 This paper points out the eXplainable Artificial Intelligence or XAI issues in software
189 development. Specifically, the authors address the issues of injecting explainability in AI
190 and ML models from the software engineering perspective. The purpose of deploying
191 software engineering framework is to assist human decision making process. It is often
192 times an ethical issue that needs to be solved instead of a technical issue. The success of
193 AI-based assistant holds its crucial role at the center of the explainability power. The
194 contemporary software development is a process of integrating, building, implementing
195 many open source projects in an organic way. This process is often times checked by
196 many software engineers while each carries a specific role and sets of knowledge to
197 develop one block of code as a black-box without many questions. The authors showed
198 that only 38% of the prediction studies came with global explanations and only 5% came
199 with local ones. These results show lack of confidence that human users may have if
200 are asked to deploy the models in production. To tackle this problem, the author used

201 LIME, which is a model-agnostic technique to explain the predictions of file-level defect
202 prediction models. LIME is a great candidate because it uses linear boundary to establish
203 local separation for explainability. The proposed approach is able to identify the tokens
204 that is used to contribute to the overall software program. The explanation also allows
205 programmers to debug more efficiently.

206 **4.4 Selected Work: Liu et al. (2021)**

207 The work by Liu et al. (2021) aims to explore and to predict tensorflow program bugs
208 in real world developmental environment. The first glimpse the title show ambition
209 and it is a big target to shoot for. Yet, it is quite necessary, because deep learning
210 forms the fundamental of Artificial Intelligence or AI for most of today's technology.
211 The most popular deep learning pipelines today are designed using TensorFlow from
212 Google Brain or PyTorch from Facebook AI Research or FAIR. These two packages
213 are well known and have very convenient APIs to design customized deep learning
214 models. This paper investigates large-scale empirical data set of more than 12,000
215 failed TensorFlow jobs and examined the bugs reported in the TensorFlow programs.
216 To tackle this problem, their work proposed a constraint-based approach to detect the
217 shape-related bugs. The authors tested their proposed methodology on 60 industrial
218 TensorFlow programs and showed that their work is efficient and effective. This branch
219 of research is quite fruitful and can deliver promising results to the programs designed
220 and developed in the industry. Because of wide variety of data sets that can appear in
221 the industry, this branch of research is very attractive to the practice in the industry.
222 However, the paper also indicated no false positives. This could pose as a threat to the
223 robustness of the performance. The trade-off between false positives and false negatives
224 is important to understand. In industrial pipeline, to ensure safety practice, it is much
225 more risk averse to have a lot of positives cases identified, but some of them are false.
226 This is not detrimental to the entire system, because the worst case scenario is the cost
227 of going through the program and there is no harm to the entire system. The other
228 case, false negatives, is much worse. If there are programs at fault and will harm the
229 system and these are the programs undetected (shown as negatives), the reality will be
230 detrimental and the result could harm the entire system.

231 **4.5 Selected Work: Sartaj (2021)**

232 The work by Sartaj (2021) has contributed a novel AI-based methodology to create test
233 case scenarios for unmanned aerial systems or UAS. The motivation is straightforward
234 for this branch of research, because the current UAS are evaluated on test case scenarios
235 written by hand on a case-by-case basis. This way the pipeline is much more transparent,
236 yet the procedure is cumbersome, inefficient, and inaccurate. The test settings are usually
237 designed to check a handful of simulators at once. The author proposed a AI-based
238 system to automate the UAS test case scenarios. The proposed technique is called
239 AITester. It utilizes AI-based testing algorithms to create, generate, and evaluate test
240 scenarios. The author shows preliminary results that the AITester is proficient at violating
241 the expected behavior by autopilot.

242 Like all other AI-based programs, there is a trade-off between accuracy and transparency.
243 This could lead to the next stage of research for AITester.

244 **5 ISSTA**

245 **5.1 Brief**

246 The International Symposium on Software Testing and Analysis or ISSTA is a famous
247 leading research symposium focusing on software engineering, testing and analysis.
248 Each year the ISSTA joins many different and diverse groups of people together from
249 academia and industry to talk and exchange novel ideas. The conference aims to
250 contribute to the community ground breaking research and new software systems that
251 can be steering the direction of software production in the future.

252 **5.2 Tracks**

253 The tracks for ISSTA are simple and straightforward, which are much more easier to
254 understand than the previous conferences. The technical papers hold the main spotlight
255 of the conference each year. The conference also has workshops and artifact evaluations
256 such as the other conferences. A unique track that other conferences do not have is the
257 tool demonstrations. In addition, the conference also allows doctorate students to share
258 their research as well.

259 **5.3 Selected Work: Marinescu et al. (2014)**

260 In practice, the software engineering production generates different level of source code
261 during its evolution across different life time. The software is stored on a repository
262 and the repository can be updated many times by a team or the project could involve
263 more than one teams to use the same repository for development. In this environment,
264 the version control and monitor serve a crucial role in this pipeline and currently the
265 pipeline is exhaustive where different versions could take a lot of attention and cost to
266 maintain. The author and his team proposed a new model called COVRIG. COVRIG is
267 a flexible infrastructure designed to run each version in the system in parallel and create
268 reports for each path that summarizes the performance metrics. It takes advantage of the
269 empirical study conducted to investigate how programs evolve throughout the processes
270 of code, tests and coverage.

271 **5.4 Selected Work: Zhang et al. (2013)**

272 Mutants in software engineering are syntactic transformations of the underlying program
273 in the test environment. These mutants act as measure to allow the programmers to
274 understand the given test at a certain measure, indicating whether there is a potential bug
275 in the program. The concept “kill” refers to a situation where one program test certain
276 results that come out to be completely different from the output in the original program.
277 Denote the original program O and the mutant program for test to be M . It is said
278 that the test T kills M if the test result using T on M is different from O . There is an
279 underlying assumption in the argument stating that the program M is supposed to be the
280 mutant of the original one, i.e. O . If this assumption is challenged, the argument would
281 have to be reconstructed. This procedure of mutant test could be extremely costly if
282 there is a large number of mutants. The author Zhang et al. (2013) proposed a family of
283 methodologies that focus on reducing the test costs to ensure quicker and more efficient
284 way of checking or potentially killing the unnecessary or incorrect mutants.

285 5.5 Selected Work: Loyola et al. (2014)

286 In the evaluation of a computer program, test data is created consisting of data inputs
287 with features called oracles. The test data is then tested using a computer program or a
288 programming system to make execution. The evaluation requires the test input data to
289 be fed into the system to create test results that are same with test oracles. If they are
290 different, then test oracles determine what is failed scenario in the final evaluation. The
291 test oracles are originally pre-defined or created by hand based on agreed upon protocols
292 amongst programmers. However, recent development has show novel direction that test
293 oracles should be tailored to particular test inputs to ensure generalization across different
294 programs and platforms. The authors seek novel direction to create automated test case
295 and hence generate meaningful test scenarios for the oracle to create and to be used as
296 the benchmark. It is believed that the testers, while using the proposed methodology, can
297 improve the program with high performance and also return meaningful and interpretable
298 results that can be customized to the real world situation. The proposed program is called
299 DoDona which has a proposed algorithm to test input dataflow recordings. The authors
300 show with empirical work of the performance of DoDona, which is able to improve the
301 findings of up to 115% and reduce the search cost by 90%.

302 References

- 303 Ali, R. H., Parlett-Pelleriti, C., and Linstead, E. (2020). Cheating death: A statistical
304 survival analysis of publicly available python projects. In *Proceedings of the 17th
305 International Conference on Mining Software Repositories*, pages 6–10.
- 306 Denaro, G., Margara, A., Pezze, M., and Vivanti, M. (2015). Dynamic data flow testing
307 of object oriented systems. In *2015 IEEE/ACM 37th IEEE International Conference
308 on Software Engineering*, volume 1, pages 947–958. IEEE.
- 309 Gopinath, R., Jensen, C., and Groce, A. (2014). Code coverage for suite evaluation
310 by developers. In *Proceedings of the 36th International Conference on Software
311 Engineering*, pages 72–82.
- 312 Huo, C. and Clause, J. (2014). Improving oracle quality by detecting brittle assertions
313 and unused inputs in tests. In *Proceedings of the 22nd ACM SIGSOFT International
314 Symposium on Foundations of Software Engineering*, pages 621–631.
- 315 Inozemtseva, L. and Holmes, R. (2014). Coverage is not strongly correlated with test
316 suite effectiveness. In *Proceedings of the 36th international conference on software
317 engineering*, pages 435–445.
- 318 Liu, C., Lu, J., Li, G., Yuan, T., Li, L., Tan, F., Yang, J., You, L., and Xue, J. (2021).
319 Detecting tensorflow program bugs in real-world industrial environment. In *2021
320 36th IEEE/ACM International Conference on Automated Software Engineering (ASE)*,
321 pages 55–66. IEEE.
- 322 Loyola, P., Staats, M., Ko, I.-Y., and Rothermel, G. (2014). Dodona: automated oracle
323 data set selection. In *Proceedings of the 2014 International Symposium on Software
324 Testing and Analysis*, pages 193–203.
- 325 Malhotra, R., Kapoor, R., Aggarwal, D., and Garg, P. (2021). Comparative study of
326 feature reduction techniques in software change prediction. In *2021 IEEE/ACM
327 18th International Conference on Mining Software Repositories (MSR)*, pages 18–28.
328 IEEE.

- 329 Marinescu, P., Hosek, P., and Cadar, C. (2014). Covrig: A framework for the analysis
330 of code, test, and coverage evolution in real software. In *Proceedings of the 2014*
331 *international symposium on software testing and analysis*, pages 93–104.
- 332 Nguyen, H. A., Dyer, R., Nguyen, T. N., and Rajan, H. (2014). Mining preconditions
333 of apis in large-scale code corpus. In *Proceedings of the 22nd ACM SIGSOFT*
334 *International Symposium on Foundations of Software Engineering*, pages 166–177.
- 335 Sartaj, H. (2021). Automated approach for system-level testing of unmanned aerial
336 systems. In *2021 36th IEEE/ACM International Conference on Automated Software*
337 *Engineering (ASE)*, pages 1069–1073. IEEE.
- 338 Shi, A., Gyori, A., Gligoric, M., Zaytsev, A., and Marinov, D. (2014). Balancing trade-
339 offs in test-suite reduction. In *Proceedings of the 22nd ACM SIGSOFT international*
340 *symposium on foundations of software engineering*, pages 246–256.
- 341 Tantithamthavorn, C. K. and Jiarapakdee, J. (2021). Explainable ai for software engi-
342 neering. In *2021 36th IEEE/ACM International Conference on Automated Software*
343 *Engineering (ASE)*, pages 1–2. IEEE.
- 344 Tufano, M., Watson, C., Bavota, G., Di Penta, M., White, M., and Poshyvanyk, D.
345 (2018). Deep learning similarities from different representations of source code. In
346 *2018 IEEE/ACM 15th International Conference on Mining Software Repositories*
347 *(MSR)*, pages 542–553. IEEE.
- 348 Zhang, L., Marinov, D., and Khurshid, S. (2013). Faster mutation testing inspired by
349 test prioritization and reduction. In *Proceedings of the 2013 International Symposium*
350 *on Software Testing and Analysis*, pages 235–245.