Bluejay: A Cross-Tooling Audit Framework For Agile Software Teams

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Abstract—Agile software teams are expected to follow a number of specific Team Practices (TPs) during each iteration, such as estimating the effort (“points”) required to complete user stories and coordinating the management of the codebase with the delivery of features. For software engineering instructors trying to teach such TPs to student teams, manually auditing teams if teams are following the TPs and improving over time is tedious, time-consuming and error-prone. It is even more difficult when those TPs involve two or more tools. For example, starting work on a feature in a project-management tool such as Pivotal Tracker should usually be followed relatively quickly by the creation of a feature branch on GitHub. Merging a feature branch on GitHub should usually be followed relatively quickly by deploying the new feature to a staging server for customer feedback. Few systems are designed specifically to audit such TPs, and existing ones, as far as we know, are limited to a single specific tool.

We present Bluejay, an open-source extensible platform that uses the APIs of multiple tools to collect raw data, synthesize it into TP measurements, and present dashboards to audit the TPs. A key insight in Bluejay’s design is that TPs can be expressed in terminology similar to that used for modeling and auditing Service Level Agreement (SLA) compliance. Bluejay therefore builds on mature tools used in that ecosystem and adapts them for describing, auditing, and reporting on TPs. Bluejay currently consumes data from five different widely-used development tools, and can be customized by connecting it to any service with a REST API.

Video showcase available at governify.io/showcase/bluejay

Index Terms—team practice, agile, team dashboard, team practice agreement

I. INTRODUCTION

Agile software development has become widespread in both industry and in the classroom. Its lightweight processes and methodologies are increasingly well supported by external tools; Figure 1 shows a typical workflow for projects in which different teams individually to a large code repository. The upstream repository is forked and each planned feature is developed by a member of the team in a separate branch. Once a feature is completed, a pull request (PR) is opened for the whole team to discuss the changes, and if accepted, merge the changes to the forked repository’s main branch. Once merged, the main branch is deployed to the staging server for the client to validate new functionality. If the client signs off on the feature, another PR is opened to merge the changes into the main repository so the feature can be included in the production server.

Many tools have emerged to support such workflows. Project management tools such as Pivotal Tracker (PT) allow users to organize tasks and track who is working on which features. Repository management tools such as GitHub can be used to manage the code and coordinate the merging of new code into the main line. Continuous integration tools such as Travis CI run tests and linters to validate changes before accepting any PR. Deployment-as-a-service platforms such as Heroku even allow automatic deployment from specific branches.

Well established metrics for Agile processes rarely capture TPs that consist of sequences of actions across such tools. For example, when following a “branch per feature” development approach, a branch should be created shortly after a feature is claimed by a developer. There exist multiple tools that inte-
grate with GitHub or PT to help audit teams such as ZenHub or Datadog but none of them are capable of correlating multiple tool information. Therefore, both for Agile teams in industry and for instructors teaching these methodologies, it is difficult to audit if teams are following practices that integrate multiple tools correctly.

Fortunately, there is already another domain in which the challenge is to calculate performance indicators from multiple services. It is in the domain of Service Level Agreement (SLA) management, and more specifically, expressing and evaluating Service Level Objectives (SLOs). We find an important similarity between a TP structure and SLO in SaaS. For example, a typical SLO for an interactive Web site might be “over any 5-minute window, 99% of requests to the main page must be delivered within 200ms”. Analogously, a typical TP for an Agile team might be “over any 2-week iteration, 75% of stories should be ‘1-point’ stories”. In such a context, a set of TP define a global agreement, that we coin as Team Practice Agreement (TPA), in a similar way that a set of SLOs define a service level agreement (SLA). Therefore we can leverage the models and tools that have already been successful for SLOs and adapt them for TPAs.

Bluejay leverages the above insight for the challenge of multi-tool TPAs. It offers a systematic way to define, measure and visualize TPAs involving up to two different services. The framework provides a tooling ecosystem for instructors to define their own best practices to follow and track the adherence of students teams in order to learn of pitfalls and improve over time.

Specifically, the framework provides a micro-services architecture based on the Governify ecosystem for SLA management. It has been extended in two different ways: a new Domain Specific Language (DSL) for adopting different tools in order to create TPAs, and a customized dynamic dashboard that shows the degree of adherence to practices over time for the different teams. We have created components that connect to popular existing Agile tools, but Bluejay is extensible both by adding new external tools and by defining new metrics and practices based on data from any supported external tools.

Section 2 introduces our system, including how to model, express, and audit Team Practice Agreements (TPAs). Section 3 describes our experience using the tool in real academic scenarios. Section 4 discusses related work. Section 5 reviews our plans for future work.

II. BLUEJAY

Bluejay is a framework composed of a set of components that allows modeling, monitoring, and auditing TPAs. In Figure 2 a general vision of the process involving Bluejay is displayed. The figure distinguishes a general workflow composed by three stages:

1) First, Modelling is the process by which Instructors define a set of TPAs, creating a TPA for teams to fulfill.
2) Second, a monitoring process for each TPA is started; as part of this process Bluejay calculate the TPA fulfillment based on the metrics computed with the information gathered from the different REST APIs; the actual fulfillment information and metrics values are subsequently rendered in a graphical dashboard available for both instructors and teams.
3) Third, Instructors can use the dashboard to audit teams and determine if they adhere to the practices.

A. TPA Modeling

The user data input in Bluejay are the Team Practice Agreements, or TPAs. A TPA is composed of multiple TPAs, each of which involves one or more metrics, and each metric is computed using the information from one or more tools.

A Metric Pattern (MP) is a template for collecting, aggregating, measuring, and evaluating a raw metric from a specific tool. An example of a raw metric might be a count from PT of the number of currently-active user stories in the team’s project (stories that have been started but not completed).

Another example might be the overall test coverage of the most recent CI run of the project, as reported by a tool such as Travis CI. We identified four Metric Patterns that capture a variety of Team Practice behaviors useful in expressing TPAs. Figure 3 shows the four metric patterns and an example of a Team Practice that uses each. We inspired in how process performance indicators are specified by using linguistic patterns as Travis CI. As the table shows, each TP is defined by a metric, an objective defining the valid value range for the metric, and a scope that determines whether the metric should be applied to the whole team or to each member separately. Each instructor determines which TPAs they wish to audit for their students. The tool’s complete documentation details which raw metrics are available, how to define new TPAs, and how to add new additional external tools’ metrics and TPAs.

The TPA must be expressed in the syntax of iAgree, a YAML-based domain-specific language used in Governify. Figure 4 shows the iAgree YAML corresponding to TP #4 in the table. The expression evaluates to the percentage of started stories in PT that are correlated in time with a branch creation...
### Metric Patterns

| Pattern Description | Example TP | Objective |
|---------------------|------------|----------|
| Number of [Event] in [Tool] every [Period] by [TEAM|MEMBER] | TP#1: Number of OPEN_PR in GITHUB every WEEK by MEMBER | <= 1 |
| [MAX|MIN|AVG|STD|NEWEST|LATEST] [Property] value of [Event] in [Tool] every [Period] by TEAM | TP#2: AVG COVERAGE value of COVERAGE_REPORT in CODE by TEAM | <= 75% |
| [Frequency] distribution of [Event] in [Tool] every [Period] by TEAM | TP#3: DAILY distribution of SUCCESSFUL_BUILDS in TRAVIS every WEEK by TEAM | σ <= 1 |
| Percentage of [Event1] in [Tool1] correlated with [Event2] in [Tool2] within [window] every [period] by TEAM | TP#4: Percentage of NEW_BRANCH (creation) in GITHUB correlated with START_STORY in PIVOTAL_TRACKER within 1_HOUR every WEEK by TEAM | >= 75% |

Fig. 3. Four metric patterns available for defining metrics of interest in Team Practices, and example Team Practices TP1–TP4 that follow each of the patterns.

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A catalogue of examples is available for public usage; later in the paper we describe how Bluejay can be extended with additional external tools, additional metrics, and additional metric patterns.

### B. TPA Monitoring

Once TPAs are defined in iAgree, Bluejay can begin monitoring them. The middle dotted box in Figure 2 shows the main components of Bluejay. The **Registry** stores TPAs in the iAgree language. The **Reporter** is the component responsible for the main workflow; it coordinates the work of **collectors**, each of which is customized to retrieve data and compute metrics for a unique external tool. A TPA and its TPs can use multiple collectors. When the TPA is fully calculated, the reporter dynamically adapts the dashboard for auditing the team. The initial implementation of Bluejay has three collectors, but more can be added:

- **PTCollector**: Collects data from PT, for defining metrics related with PT stories for the team or for each individual member. E.g. “One story started at a time for each member”
- **GHCollector**: Collects data from GitHub for defining metrics related to GitHub Projects boards. E.g. “One card in "In progress" column at a time”.
- **EventCollector**: A generic way to fetch data from 2 different APIs and correlate two different payloads. This is used to create complex TPAs that more accurately represent the needs of auditing Agile teams using multi-tool workflows. Currently, the EventCollector is able to connect with 5 different data sources: PT, GitHub, Heroku, TravisCI and Codeclimate.

### C. TPA Auditing

Bluejay offers diagrams to audit both team-level TPAs and member-level TPAs. The system will automatically generate a dashboard for each TPA based on the scope of each guarantee. Both graphs will appear as a pair of graph and table, containing the obtained values after computing all the guarantees. The X axis of the graph corresponds to the time and the Y axis to the actual value of the metric. The table contains the same points but displayed as a pair of date and value. Figure 5 shows a screenshot of a fragment of the dashboard including a graph representing TP#4 from figure 3.

The top (team) graph in the figure shows that most of the time, the team creates a new feature branch in GitHub within 1 hour of starting a story on PT. On the right side, the actual value of the points for each date can be seen inside a green box if that point fulfills the objective or a red box if not. The bottom (team members) graph shows a colored line for each team member, and the table at the top shows each member's compliance with TP#1, which may be stated as “Each member should open at least one PR every week.” Users can change the period of time to analyze. This is useful when looking for practice violations within a certain iteration.

### D. Customization and Extension

The patterns described in the second section categorize the metrics that can be used currently in the system. The system is prepared to be extended in five ways:

- **TPAs** can be created and extended to audit all the TPAs certain team needs. There is an available catalogue of examples to use as starting point.
- **Metric**: As the Event Collector can combine information from multiple tools or extract information in a generic

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PERCENTAGE_NEWBRANCH_STARTEDSTORIES:

```yaml
  measure: actual
  element:
    percentage: related
github:
  events:
    type: CreateEvent
    payload:
      ref_type: branch
      window: 3600
  event:
    pivotal:
      activity: started
```

Fig. 4. iAgree YAML syntax corresponding to TP#4 in Figure 3.
As an adaptation example, if a instructor needs to integrate Trello for auditing teams, if the metric follows any pattern of the presented in Figure 3, the only requirement is to extend the event collector to include the new data sources for the Trello API. If the metric is not compatible with the given patterns there are two options: extending the event collector to support the new metric, or creating a new collector particularly for the specific type of metric.
**Findings.** Students reported better backlog delivery behaviors in iteration 3 and 4. The tool reported better adherence to the practice during iteration 2 when the TPs where given to the students. During the 3rd iteration there was a small reduction in adherence, which could be caused by a fatigue phenomenon. Finally in the iteration 4 the tool reported the highest adherence, when the dashboard was available to students.

**Summer 2020.** We analyzed projects from a Agile software engineering summer course in 2020. Groups of one, two or three students formed a total of 31 teams. Teams had a single ten-day sprint to complete all the features. We examined if students followed each of the following three pairs of events

- How often does starting a story correspond with creating a branch?
- How often does finishing a story correspond with opening a PR?
- How often does delivering a story correspond with merging a PR?

Each team had a Github repository and a Pivotal Tracker project to track progress on the features, and both services were audited with Bluejay. They also were taught to integrate GitHub and PT in order to correlate those events.

**Findings.** In the ideal version of our workflow, each story is associated with exactly one branch, the branch is created when work on the story is started and the branch is merged when the story is delivered. In the practice we found that this correspondence is not always one-to-one and some teams did not connect their “started” stories to its branch; also, many stories that were marked as “finished” but not associated with an opened PR possibly due to a number of reasons: they never connect the story with a branch so there wouldn’t be one to find a PR from, or, a branch might have been merged without opening a PR to solicit review. Finally, many stories were marked as “delivered” but were not connected with a merged (closed) PR. This could be for at least two reasons. First, “delivered” stories may have had PRs opened for them that could not be merged before the course ended (students ran out of time), and second, for reasons similar to “finished” stories not being connected with open PR.

**IV. RELATED WORK**

To the best of our knowledge, currently no related tooling has been proposed in the literature. In [8], we proposed an initial prototype of the system for auditing team practices but it only allow the usage of metrics in an isolated context lacking the cross-tooling correlation (e.g. TP#4) that represents a fundamental feature to analyse the flow of the overall process.

However, related work around professional software teams defines software development practices that could be defined as TPs in order to be measured and analyzed with our proposed tool. Thus, Baltes et al. propose in [9] a theory describing software developers’ properties and practices that are distinctive for experts in their field. For example, the concept of getting feedback from peers is identified as an important factor that could be used as a metric in our TPs. Huijgens et al. [10] provide metrics with high predictive power towards a subset of lagging variables. Thus, as part of their results, metrics such as “planned stories completion ratio” or “planned points completion ratio” could be defined as TPs in our work, as the authors ascribe to them a high predictive power. Treude et al. [11] and Meyer et al. [12] conducted empirical studies with 156 GitHub users and 379 software developers. From their studies metrics emerged for both (1) analysis for objective measures of development activity and (2) improvement of productivity through the development of new tools and the sharing of best practices. Bluejay could be extended in this direction to incorporate such a set of new metrics. Finally, Kupiainen et al. [13] establish a motivation for our work by providing a systematic literature review analyzing why and how metrics are used by industrial agile teams in order to infer enforcing process improvements.

In software engineering education, there are recent studies on exploiting metrics and dashboards to provide fast feedback. Matthes et al. [14] show how to use a metric dashboard, ScrumLint, to provide fast feedback to students. Bai et al. use metrics to deliver continuous feedback to students in a software engineering course [15]. Our approach generalizes these techniques and provides an extensible framework that integrates with external tools for capturing these and other Team Practice metrics that instructors design.

One challenge instructors of project-based Agile courses face is ensuring that students gain an understanding of core Agile concepts. Steghöfer et al. [16] found that students felt delivering a final product was more important than understanding Agile processes. Bluejay can help instructors measure student adherence to Agile processes by describing the Agile practices taught in class as TPAs. Based on the scores for each TPA, an instructor can quickly come to a decision about which Agile practices are not being followed and intervene with the appropriate student or team.

**V. CONCLUSIONS AND FUTURE WORK**

Bluejay leverages the existing Service Level Agreement (SLA) management platform Governify [1], extended with a DSL for Agile software development metrics and an interactive dynamic dashboard. The result is automated auditing of team practices widely used in agile methodologies, especially when those practices involve computing metrics from data aggregated from multiple tools supporting Agile workflows. As future work we plan to extend the infrastructure adding more tools and metric patterns, and to continue validating the system in software teams where we can mine interesting TPs from historic data and also audit new teams.

Beyond an Agile context, Bluejay opens the possibility to a more systematic approach for supporting software engineering teaching, first with a way of defining a set of team practices for the teams to follow and also when observing the results archived for a continuous improvement. It is not easy to do this process manually and our tool is able to automate it. Bluejay can also be used for software engineering teaching experimentation, for example, defining a set of potential best
practices and then observing how teams’ results are affected if they follow these practices.

More information and installation instructions for Bluejay may be found at: https://governify.io/quickstart/auditing-agile

REFERENCES

[1] C. Müller, P. Fernandez, A. M. Gutierrez, O. Martín-Díaz, M. Resinas, and A. Ruiz-Cortés, “Automated validation of compensable slas,” IEEE Transactions on Services Computing, 2018.

[2] A. del Río-Ortega, M. Resinas, A. Durán, and A. Ruiz-Cortés, “Using templates and linguistic patterns to define process performance indicators,” Enterprise Information Systems, vol. 10:2, pp. 159–192, 2016.

[3] G. César, G. Alejandro, Z. Joshua, S. Kollakunta, F. Pablo, F. Armando, and A. Ruiz-Cortés, Bluejay: Metrics’ catalogue. [Online]. Available: https://github.com/isa-group/governify-examples/tree/master/metrics/event-collector

[4] ———. Bluejay: Tpas’ catalogue. [Online]. Available: https://github.com/isa-group/governify-examples/tree/master/TPAs/bluejay

[5] ———. Bluejay: How to extend event collector’s sources. [Online]. Available: https://github.com/isa-group/governify-collector-events/wiki/How-to-increment-collector-with-new-API

[6] ———. Bluejay: Collector template. [Online]. Available: https://github.com/isa-group/governify-collector-v2-template

[7] ———. Bluejay: Custom graphs. [Online]. Available: https://governify.io/customization/dashboards

[8] A. Guerrero, R. Fresno, A. Ju, A. Fox, P. Fernandez, C. Muller, and A. Ruiz-Cortés, “Eagle: A team practices audit framework for agile software development,” in Proceedings of the 2019 27th ACM Joint Meeting on European Software Engineering Conference and Symposium on the Foundations of Software Engineering, ser. ESEC/FSE 2019. New York, NY, USA: ACM, 2019, p. 1139–1143. [Online]. Available: https://doi.org/10.1145/3338906.3341181

[9] S. Baltes and S. Diehl, “Towards a theory of software development expertise,” in Proceedings of the 2018 26th ACM Joint Meeting on European Software Engineering Conference and Symposium on the Foundations of Software Engineering, ser. ESEC/FSE 2018. New York, NY, USA: ACM, 2018, pp. 187–200. [Online]. Available: http://doi.acm.org/10.1145/3236024.3236061

[10] H. Huijgens, R. Lamping, D. Stevens, H. Rothengatter, G. Gousios, and D. Roman, “Strong agile metrics: Mining log data to determine predictive power of software metrics for continuous delivery teams,” in Proceedings of the 2017 11th Joint Meeting on Foundations of Software Engineering, ser. ESEC/FSE 2017. New York, NY, USA: ACM, 2017, pp. 866–871. [Online]. Available: http://doi.acm.org/10.1145/3106237.3117779

[11] C. Treude, F. Figueira Filho, and U. Kulesza, “Summarizing and measuring development activity,” in Proceedings of the 2015 10th Joint Meeting on Foundations of Software Engineering, ser. ESEC/FSE 2015. New York, NY, USA: ACM, 2015, pp. 625–636. [Online]. Available: http://doi.acm.org/10.1145/2786805.2786827

[12] A. N. Meyer, T. Fritz, G. C. Murphy, and T. Zimmermann, “Software developers’ perceptions of productivity,” in Proceedings of the 22Nd ACM SIGSOFT International Symposium on Foundations of Software Engineering, ser. FSE 2014. New York, NY, USA: ACM, 2014, pp. 19–29. [Online]. Available: http://doi.acm.org/10.1145/2635868.2635892

[13] E. Kupiainen, M. V. Mäntylä, and J. Ikonen, “Why are industrial agile teams using metrics and how do they use them?” in Proceedings of the 5th International Workshop on Emerging Trends in Software Metrics, ser. WETSoM 2014. New York, NY, USA: ACM, 2014, pp. 23–29. [Online]. Available: http://doi.acm.org/10.1145/2593868.2593873

[14] C. Matthies, T. Kowark, K. Richly, M. Uflacker, and H. Plattner, “How surveys, tutors and software help to assess scrum adoption in a classroom software engineering project,” in 2016 IEEE/ACM 38th International Conference on Software Engineering Companion (ICSE-C). IEEE, 2016, pp. 313–322.

[15] X. Bai, M. Li, D. Pei, S. Li, and D. Ye, “Continuous delivery of personalized assessment and feedback in agile software engineering projects,” in 2018 IEEE/ACM 40th International Conference on Software Engineering: Software Engineering Education and Training (ICSE-SEET). IEEE, 2018, pp. 58–67.

[16] J. Stegöhöfer, E. Knauss, A. Alégroth, I. Hammouda, H. Burden, and M. Ericsson, “Teaching agile - addressing the conflict between project delivery and application of agile methods,” in 2016 IEEE/ACM 38th International Conference on Software Engineering Companion (ICSE-C), 2016, pp. 303–312.