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(Full Paper)

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ABSTRACT
Many open source software (OSS) project creators adopt GitHub as their chosen online repository. They seek out others within the global OSS community of developers. Such community developers are then encouraged to add their capabilities, ideas and coding into a creator's developing OSS project. A structural equation modelling study of three top OSS programming languages deploys GitHub's operational elements as a four stage directional suite of (1) dependent, (2) intermediaries, and (3) independent elements. It shows a project's activity levels can be enhanced when additional project contributions are effectively stage-wise pursued. A staged development approach helps creators understand the process of attracting OSS developers into a creator's GitHub project.

Keywords: GitHub, open source software; OSS, developer repo, big data

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INTRODUCTION AND MOTIVATION
GitHub is the most popular hosting site for open source software (OSS) development and repositories (Cosentino, Izquierdo & Cabot, 2017). It houses over 57 million project repositories, with contributions from over 28 million developers (GitHub, 2018). GitHub's active big data projects are places where its software developers, and/or its responders or contributors alter and generally improve their developing OSS project (Gousios, *et al.*, 2014). Occasionally, apparent GitHub projects (each housed in its own repository or repo) can just be an individual programmer's additional code storage site, or they can be a cache that stores a personal code such as a 'code cracking tool' allowing unauthorized access into an existing commercial software package such as illegal pathway into the Chinese internet..

Over time, GitHub's active OSS developer repositories grow in size. Here, new knowledge and additional software capabilities are included as: (1) project corrections and improvements are made, (2) issues are discovered, (3) code is tested, stabilized and solved, (4) observers and raters offer the project external recognition, (5) coders, external developers, and contributors enhance the project's capabilities, and (6) new code, ideas, and documentation are accepted into a specific GitHub OSS project's master branch. Thus, a GitHib repository is dynamic ecosystem (Alshomali *et al.*, 2017) made up of ongoing contributing elements built by a global suite of uniquely-skilled, code-related contributors (Tsay, Dabbish & Herbsleb, 2014).

As little is known of how a GitHub ecosystem of contributors interact, this pilot study models the three most frequently engaged GitHub programming languages. It tests whether GitHub's operational elements do actually combine and align into a directional suite of (1) dependent, (2) intermediary, and (3) independent elements. It also seeks answers as to whether these combined elements exert an effect (or effects) that can provide other repo developers with a potential way to speed their individual project's 24/7/365 net activity level developments.

BUILDING THE GITHUB ECOSYSTEM ELEMENT MEASURES
Theoretical Basis
Within GitHub's OSS platform, teams of software developers act as communities and associates. They combine their interests into solving problems and they build operational or release versions of software (Wu *et al.*, 2014). This behavior fits across 'Information Integration Theory' which also builds on the common beliefs and behaviors of the 'Theory of Planned Behavior' (Ajzen, 1991). It links each community of GitHub software developers' attitudes with their subjective norms (Ajzen, 1991) in working collectively, and intentionally, on a project. Thus, a behavioral strength-of-belief is generated, and together they theoretically combine to jointly create an increase in a project's activity level. This also likely reduces this project's overall OSS development time.

Further theoretical support emerges through the 'Theory of Social Translucence' which draws a broad, visual, community-wide awareness around a project, its design strategies, and its deliverable(s) targets (Erickson & Kellogg, 2000). Here, each member of the project's community is recognized for their contributions, and ongoing activities – again ensuring collaboration into and

throughout each shared project. (Dabbish *et al.*, 2012) also support that such community behaviors do occur within GitHub projects.

Natural Ecosystem theory infuses a living community with its non-living organisms, whilst social network theory understands the interactions of participants inside a network. These theories jointly underpin that each element interacts across biological, physical and chemical processes. Similarly, in a GitHub repo ecosystem, the living community of OSS developers, observers, and interested parties interact within the GitHub project, and together they build an operational software solution that runs on a non-living hardware platform. A GitHub ecosystem also generates participant gratification which may be further behaviorally-understood through user gratification theory and motivation-consumption-gratification theory. (Katz, 1959; Severin & Tankard, 2000; Oliver & Raney, 2011; Hamilton & Tee, 2016).

Research Agenda
Hence, this study considers both such theory, and the behavior of OSS developers contributing into 600 top GitHub language specific projects. It study's 200 projects for the three most popular GitHub programing languages, and engages their OSS element contributions to poses answers to the research questions:

- ▪ *RQ1: do GitHub's operational elements actually combine and align into a directional suite of (1) dependent, (2) intermediary, and (3) independent elements for a programming language?*
- ▪ *RQ2: do these combined elements exert an effect (or effects) that can provide other repo developers with a potential way to speed their individual project's net activity levels?*

GITHUB ECOSYSTEM CONTRIBUTING ELEMENTS

A GitHub project commences when a project creator seeks OSS development assistance to create a repo as generically shown in Figure 1. Those on social media, hacker sites, or on GitHub observation watch the project. There are both passive OSS watchers who merely observe a GitHub repo, and active OSS developer watchers who monitor and occasionally comment on this GitHub repo. These constitute the 'watchers' element. Others note the quality or direction of the project and rate it. These represent the 'stars' element. Still others take a sample of the project creator's code and either use it themselves for other purposes, or develop it. These skilled OSSS developers make-up the 'forks' element. These three elements collectively engage and commence to assist the initial creator's project. Hence these independent variables constitute the initial engagement (or interest) group.

Figure 1: Aspects of a GitHub Repo

Table 1 summarizes the GitHub open source developer contributions to the above repo.

Table 1 Repo contributions

Repo Encounters	Description
Watchers	developers who observe rrepo content changes
Stars	developers who indicte a liking for the repo
Forks	developers who take a copy of the repos to work on
Issues	developers who solve a software issue for the repo
Pulls	developers who feed-back a potential solution to improve the repo
Contributors	developers who build a direct contribution to the repo
Releases	milesstone in the development life cycle of the repo
Commits	new addition/update to the current repo

When a developer clones a repo, this gives them a complete copy of the repo content without necessarily being an active part of that repo. Unfortunately, GitHub statistics do not track information about cloning. When a fork element is then feedback (with new developments) into the creator's original project, and these points are accepted, then this action contributes a 'pull' element. An OSS developer can also recognize a problem that need to be solved (arising from several possible sources). This constitutes an 'issues' element, which once solved is included in the creator's OSS improvements. As time progresses a 'release' element or new, improved version of the creator's OSS code may emerge. As OSS developers come to the creator's project, some skilled ones are invited to join as new additions to the 'contributors' element. Thus the intermediate variables may split into two groups – one raising expectations, and the other adding values to the project.

All these community elements assist the creator in the build of a successful and useful OSS outcome. Each addition that is committed to the project is termed a 'commit' element. Hence the more accepted commits, the higher the project's activity level. This dependent variable (commits) delivers levels of gratification to its OSS community.

Thus, the GitHub project repo has at least three main community blocks of OSS elements: (1) a starting, dependent, engagement/interest group, (2) an intermediary, participant and solvers group, and (3) an intermediary values capturing group, and (4) an independent or OSS activities outcome group. We conceptually model these elements as Figure 2.

GitHub offers select approaches when analyzing its data. However, to build a popular GitHub project, a forks approach is typically selected. Aggarwal, Hindle & Stroulia, (2014) suggest extracting top repos against number of forks generally offers clearer, more-consistent documentation advice, and this helps to draw-in other coding contributors (Hata *et al.*, 2015). Documentation is also related to issues, solutions, and testing (Tsay, Dabbish & Herbsleb, 2014) and to social media and websites (Jiang *et al.*, 2017), again suggesting forks, watchers and stars are also likely related.
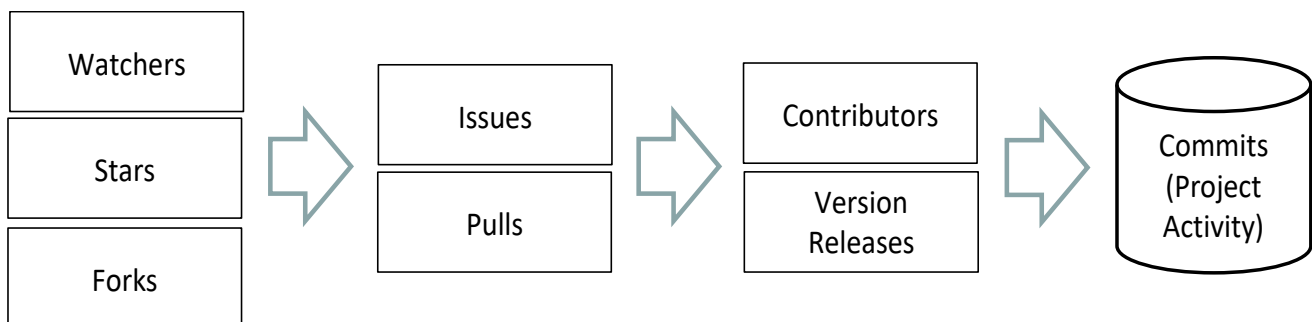

Figure 2: GitHub project ecosystem model

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METHODOLOGY

This study extends and refines Alshomali *et al.*'s (2017) and Hamilton *et al.*'s (2017) papers. It engages three programming languages, and follows a similar procedure for data capture. It requires many weeks of continuous downloading to collect the programming language repo data sets. Hence, the number of top repos for each of the top three GitHub programming languages engaged in this study is limited at 200 repos each. This number of repos per language allows for a more definitive structural equation model finding, as with over 20 cases per model construct a valid model can be established for each programming language studied (Hair *et al.* 2010).

Data Capture
This study again uses GitHub's web-based API. It again extracts GitHub data using this team's developed code (available at https://github.com/ozyjay/GithubQuery). This process of data extraction takes weeks of continual download time. Captured data is then collated into eight constructs with pulls open and closed combined into a pulls construct, and issues open and closed combined into an issues construct. Eight GitHub constructs and their measures are then imported to SPSS and AMOS 23.0 and SEM path modelled.

Data Analysis
The data is visually assessed to remove any outlier cases such as individual code storage cases, or any anti-social/'crack code tool' cases, or any non-release sites from each programming language's data set. This discriminant validity check enhances the accuracy of the targeted data and its modelling.

Table 2. GitHub construct elements (top 200 projects per language)

| GitHub Elements	Watchers	Stars	Forks	Releases	Contributors	Issues	Pulls	Commits	Total
JavaScript	165153	3553983	809328	15324	46707	434580	241377	708605	5,975,057
Python	100346	1551498	426286	10897	44978	296256	306725	1178386	3,915,372
Java	119342	1495595	527948	9230	14563	185634	130475	941579	3,424,366
									13,314,795

For these top 200 active projects in JavaScript, Python and Java the total element measures total 13.3 million, and all relate to project code or information contributions. Each project is current, active, and has existed in GitHub for at least one year. The dependent outcome variable commits is measured as the number of accepted changes into the repo of the project.

In this study we expect path differences, as OSS developers working in one of these three languages likely bring different individual talents and code building capabilities to their chosen OSS project. Also, JavaScript and Python are more a client side scripting language and Java is more a server side language activates web/file targets (and runs fast across platforms). Thus this study deploys different languages and it looks for commonalities across Figure 2's elements.

Each GitHub OSS project activity level structural path model, for each of the three GitHub OSS programming languages, is shown below across Figures 2 to 4. These three Figures each engage their respective Table 2 programming language data sets.

After outlier removal (typically those projects that: (1) are not actually OSS developer community projects, or (2) involve anti-social activities, or (3) have just one developer, or (4) show no releases) all three programming language SEM path models show excellent fit.
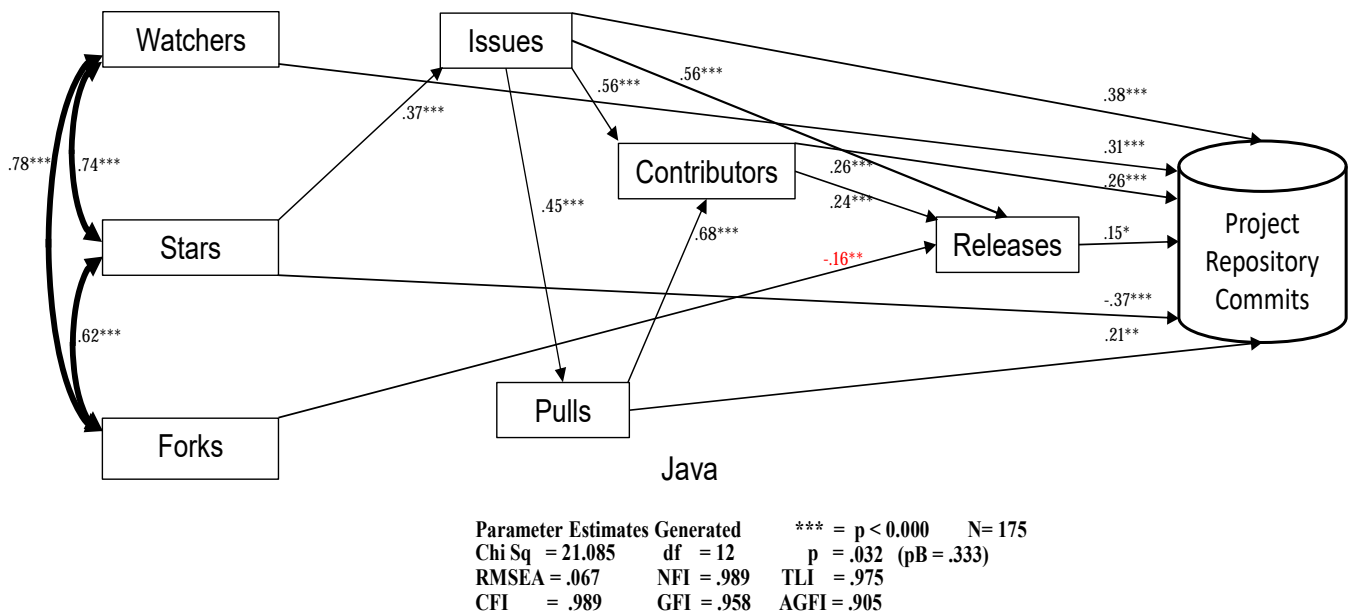
The $\chi 2/df$ ratios (between 1 and 3) supported by p values (>0.05) indicate excellent fit. All other measures displayed also support excellence of fit. (Cunningham, 2008; Hair *et al.*, 2010). All models have data sets in excess of 160 cases (20 per element) (Hair *et al.*, 2010; Muthén & Muthén, 2002). The maximum number of additive paths towards the dependent outcome variable is five. This is an acceptable maximum structural path length (Hair *et al.*, 2010). Validation by bootstrapping (200 times) further confirms model convergence and validity (Cunningham, 2008; Hair *et al.*, 2010). Thus each structural path model is representative of its top OSS GitHub programming language, and each offers suitable representative insights into the stage-wise behavior of their constituent elements.

JavaScript
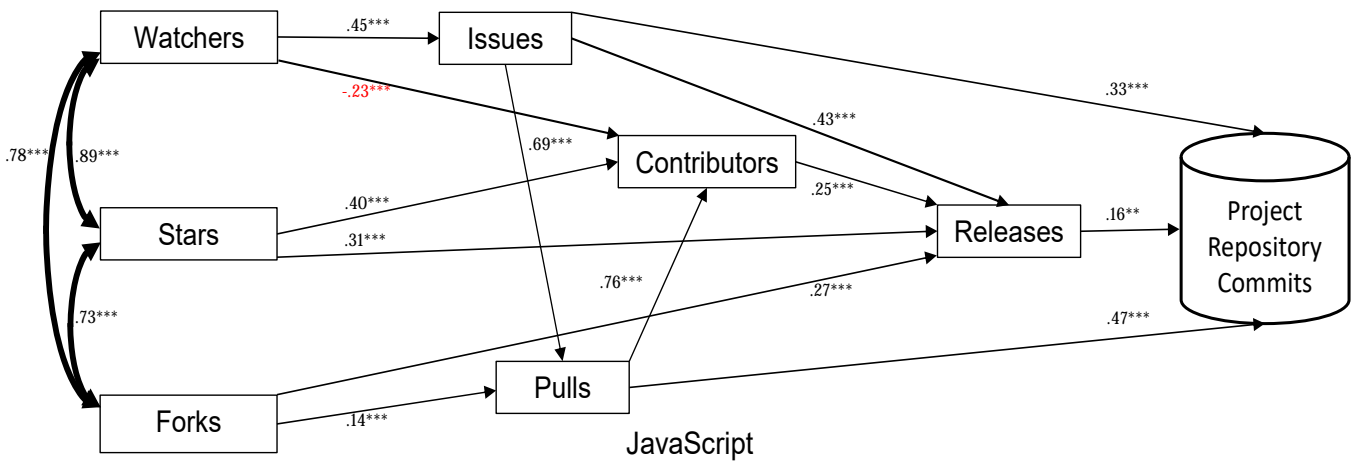
Parameter Estimates Generated   ***** = p < 0.000**   **N = 195**
Chi Sq = 20.143   **df = 12**   **p = .064**
RMSEA = .059   **NFI = .989**   **TLI = .985**
CFI = .994   **GFI = .976**   **AGFI = .927**

Python
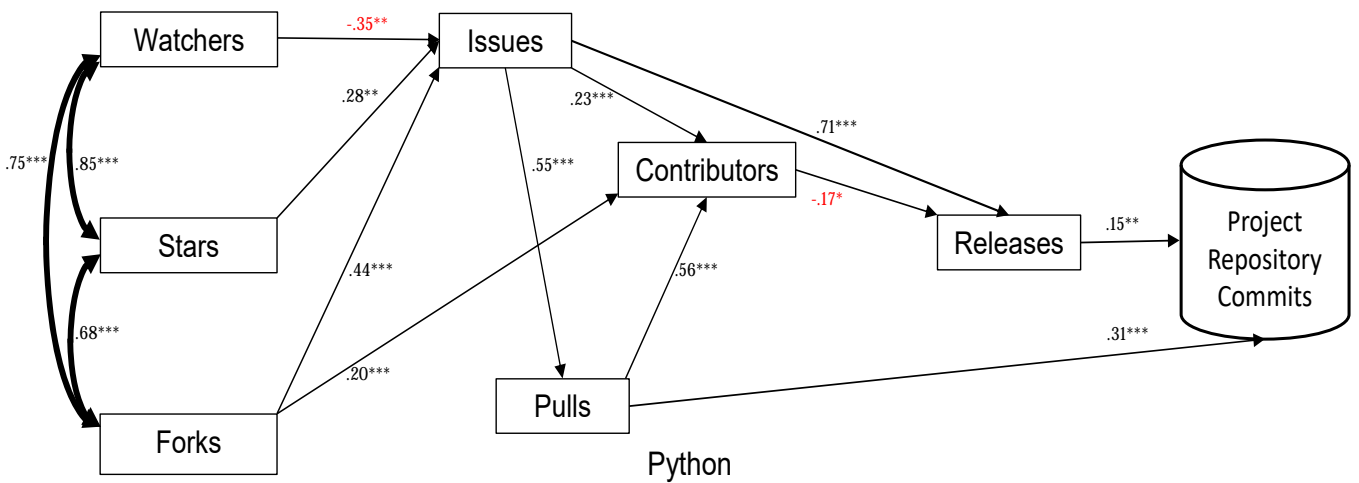
Parameter Estimates Generated   ***** = p < 0.000**   **N = 195**
Chi Sq = 21.774   **df = 14**   **p = .083**
RMSEA = .054   **NFI = .989**   **TLI = .985**
CFI = .993   **GFI = .974**   **AGFI = .932**

Java

Parameter Estimates Generated   *** = p < 0.000   N= 175
Chi Sq = 21.085   df   = 12   p   = .032 (pB = .333)
RMSEA = .067   NFI = .989   TLI   = .975
CFI   = .989   GFI = .958   AGFI = .905

Figures 3 to 5: GitHub JavaScript SEM path model for project activity level.

DISCUSSION

Figures 3 to 5 show the concept model's 4 stage transition to delivering project activity. In each case there is an internal OSS development transition from issues to pulls to contributors to releases. In each case these paths are strong and significant. Hence, based on the similarities across our three programming languages, we propose Figure 2's conceptual model as a general project activity model for GitHub. We extend this into Figure 6, indicating the likely progression of successive element influence triggers.

Considering the independent external factors (watchers stars and forks), we note watchers just track the project's activities, and generally, they do not contribute. Hence, they generally exert an indirect, and negative effect, on the overall project activity levels, Thus, although watchers and fairly highly correlated with stars and forks, a strategy to move watchers into being stars providers, or into taking forks, or even into being contributors, or into generating issues contributions is desirable when pursuing accelerated GitHub OSS project developments.

This study notes there are four negative, but significant, paths (one or two in each programming language). Each negative path is both logical for the programming language's normal use applications, and readily explained. Hence when combined with other path strength measures, each programming language SEM model is accepted as valid.
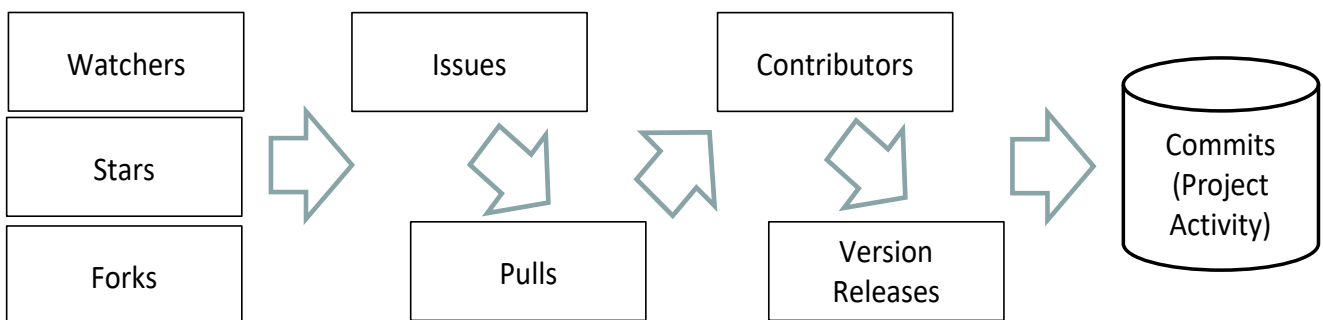

Figure 6: Progression of successive element influence triggers across the GitHub project ecosystem model

CONCLUSION

This SEM study shows that GitHub's operational elements do actually combine and align into a directional suite of (1) dependent, (2) intermediaries, and (3) independent elements for each programming language studied in GitHub. Figures 3 to 5 support the GitHub project OSS contribution elements of a project's activity level into its repository storage as a GitHub ecosystem.

The path models (Figures 2 to 4) indicate strong interrelationships amongst the elements. Thus, any improvement to the elements likely positively accelerates the GitHub projects OSS development. The SEM study also shows a consistency of combined elements exerting positive directional pathways effects towards the delivery of a project's activity levels. This is summarized in Figure 5's progression of successive element influence triggers across the GitHub project ecosystem model. This suggests a project's activity levels can be enhanced when additional project contributions are effectively stage-wise pursued. This staged development approach helps creators understand the process of attracting further OSS developers into a GitHub project.

Beyond the three languages studied, GitHub contains an additional 150+ project programming languages. Hence, this study's GitHub project ecosystem model likely has application to many of these additional languages – especially when their creators wish to (1) identify the interrelationships within each GitHub project's repo ecosystem, and/or (2) model the elements of these interrelationships as a four stage structural model, and/or (3) stage-wise attempt to accelerate the element uptakes within each of these projects, and/or accelerate their projects towards completion.

The knowledge flows embedded in this study's GitHub project ecosystem model (Figures 1 and 5) likely extends and hold ideas application to other OSS hosting platforms beyond GitHub. Those individual OSS platform creators, working on such hosting platforms, can use this study's approach when considering pathways towards specifically minimizing the OSS development timeframes involved whilst pursuing their next OSS project release.

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