

# Comparative API Complexity Analysis of Two Platforms for Networked Multiplayer Games Using a Reference Game

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**Abstract**—In this paper we propose the quantitative analysis of the complexity of a simple reference game implemented on a particular gaming platform as means for characterizing how the platform succeeds in easing the development of networked multiplayer games. We first present our own open source tool based on Sneed’s Object-Point (OP) method for the automatic quantitative assessment of the complexity of a software API by analyzing a source code using the API. We then apply our tool, together with the recently released JSComplexity tool based on classical software complexity metrics, to compare two platforms for networked multiplayer games, the open source realXtend Tundra SDK and the proprietary Union. As the reference games we use existing implementations of the simple Pong game atop the two platforms. Our data shows that these complexity metrics reveal API design tradeoffs, resulting in complexity differences in the reference games.

## I. INTRODUCTION

Reusable software libraries and platforms are pivotal for managing complexity in game development. Their APIs virtualize underlying functionality, including but not limited to: game entity management, scene construction, mathematics, input events, rendering graphics, gui input widgets, asset retrieval, game data networking, user authentication and data storage. Thus, APIs are essential for productive game development. It has been noted how making good APIs is hard - and that creating a bad API is easy [1]. Even a small quirk in an API can accumulate to substantial problems in larger bodies of application source code. API design has a significant impact on software quality, and increased API complexity is associated with increased software failure rates [2].

Also single player games encompass rich multifaceted functionality, but networked multiplayer games are typically much more complex. In network programming the developer needs to deal with events, conflicts and error conditions originating from other parts of the distributed system as well as the local user. This is emphasized in multiuser real-time systems when compared to the relatively leisurely request-response interaction patterns of most client-server applications. Developing a massively multiplayer online game (MMOG) has been estimated to typically take two to three times longer than creating and launching a single-player game [3].

Rigorous quantitative assessment of a gaming platform in terms of ease and productivity of development is very challenging. Hsiao and Yuan [3] proposed four essential “ease of”

requirements for a MMOG middleware: ease of development, deployment, maintenance and change. However, they noted the difficulty of quantitatively measuring the ease of development or change, and focused only on platform scalability in their evaluation.

We propose to use the implementations of reference game(s) as means for comparing the “ease of” dimension of game APIs. In this article we examine how the quantitative analysis of a reference game implemented atop a particular game development platform enables us to characterize how the platform succeeds in reducing development complexity. Thus, we need two things: a method for quantitative analysis of software complexity, and reference game(s). The collection of reference games effectively mimics a common corpus that for example in the information retrieval research is often used to compare the performance of different information retrieval algorithms.

Inspired by recent progress in quantitative techniques for software complexity analysis [2], [5], we first propose a tool based on Sneed’s Object-Point (OP) method for the quantitative assessment of the complexity of a software API based on the automatic analysis of a source code using the API. Then we use the tool to compare Tundra’s API complexity to that of the Union platform with pre-existing implementations of the Pong game on the two platforms, i.e. Pong serves as the minimal reference implementation of a networked multiplayer game in our study. As a complementary API complexity analysis method we use the recently introduced JSComplexity tool based on classical software complexity metrics [6]. Our study shows that the API complexity metrics calculated from the reference games with the tools indeed successfully quantify the different amounts of complexity that the game developer has to manage on these two alternate platforms.

Our underlying motivation is understanding how the API of the open source realXtend Tundra SDK, recently introduced by Alatalo [4], succeeds in hiding the complexity of networked multiplayer games. We plan to use the insight and the complexity measures in on-going development of the platform and an even friendlier API.

This paper is organized as follows. After briefly reviewing related work on API complexity analysis we present our tool. Then we report our case study on using the tool to compare

the API complexity of the two platforms in the development of the Pong game. We conclude the paper with a discussion on various aspects of the study.

## II. RELATED WORK ON API COMPLEXITY ANALYSIS

A wide range of qualitative and quantitative approaches have been proposed for assessing the complexity of software APIs. The API usability research has adapted traditional qualitative usability evaluation methods of the HCI field such as thinking aloud, heuristic evaluation and cognitive walkthroughs into evaluating APIs [7]. Employed metrics include the completion times of predefined tasks, for example.

In comparison to task-based usability evaluation of GUIs, API evaluation based on human observations is challenging. When developing a small application can take even weeks, fitting valid evaluation tasks into an observation session lasting typically 1-2 hours is difficult [8]. The recently introduced peer review [9] and the concept maps [8] methods address this problem by involving real world usage of an API over a long period of time. However, they are still considerably laborious and yield only qualitative findings.

Recently, statistical (quantitative) methods have been proposed for analyzing the complexity of software APIs. Cataldo and de Sousa [2] studied two large corporate software projects and nine open source projects and found a link between the API complexity and the failure proneness of the software quantified by the number of bug reports from the field. They quantified the complexity of an API by simply calculating the number of public methods and attributes. The approach is subject to severe limitations as it fails to take into account pre- and post-invocation assumptions and possible invocation sequences.

de Souza and Bentolila [10] introduced the Metrix tool that evaluates an API by calculating Bandi's software metrics for interface size, interaction level and operation argument complexity from the API specification. After analysing eleven different APIs with the Metrix tool, de Souza and Bentolila concluded that calculating such simple metrics directly from the API specification can produce misleading results. A more complete API may appear more complex, even if it provides good abstractions that allow completing a particular task at hand with only a small subset of the API.

An alternative to evaluating the complexity of an API with simple metrics calculated from the API specification is to assess the complexity of programs developed using the API. Sobernig et al. [5] proposed exactly this in their comparative study of the complexity of four different API designs. To characterize the complexity of a particular API, they applied Sneed's [11] Object-Point (OP) analysis to software realized atop the API. The key is to apply a surrogate measurement so that a program developed using an API is analyzed, not the API specification itself.

Osmani[12] published TodoMVC, a collection of "To Do" web application implementations built with various Javascript web frameworks. Here the source codes of the minimal implementations are meant for people to read for comparative

evaluations. We propose the same approach for networked games but combined with automated quantitative analysis.

## III. API COMPLEXITY ANALYSIS TOOLS

A tool facilitating automated and quantitative analysis of the complexity of an API from an existing body of source code developed using the API would provide several advantages. First, real world data, i.e. source codes of existing applications, could be utilized. Second, in comparison to manual methods, the analysis would be faster with immediate feedback and would require much smaller amount of human labor. Third, longitudinal studies of API development would be straightforward to conduct by analyzing successive software versions. A fully automated analysis could be embedded into the continuous integration of a software bundle, to characterize the evolution of its complexity over time.

### A. Our Tool Based on Sneed's Object-Point (OP) Method

Following the work of Sobernig et al. [5], we base our tool on Sneed's Object-Point (OP) method [11]. However, in contrast to their manual data collection, our tool automatically extracts the data needed by the OP method from a program's source code. The OP method uses intermediate UML models as the data to compare programs in different languages. Importantly, the OP method allows direct tracking between indicator values and program structures, which is elementary in evaluating API designs. For example, if many codebases get a high proportion of their complexity value due to a specific part of an API, it can then be examined qualitatively.

So called API hotspots and coldspots have been previously automatically mined from source code [13]. There, however, the specific parts of an API are not analyzed as sources of complexity, but simply to identify how much they are used. Their source code mining is similar to our tool that also needs to identify which functions are called and how often to employ the OP method.

1) *Sneed's Object-Point (OP) Method:* Although the OP method was originally developed for deriving early work estimates from UML design diagrams, recently it has also been applied to analyzing the complexity and cost of existing software implementations. While the early COCOMO software cost models used simply program size (LOC, lines of code) to estimate development effort, later the more versatile Function-Point, the Data-Point and finally the Object-Point methods have emerged to incorporate functionality and other program properties into the cost estimate [14].

Sobernig et al. illustrated the relative robustness of the OP method to the simple LoC metric on two software implementations. The first software had only 48 LoC but resulted in 356.34 OP. The second software had 144 LoC but only 266.76 OP. Their reasoning was: "an API user is only exposed to an API feature chunk of low structural complexity", as "the chunks size is limited in terms of participating classes and the smallest number of operations per class" and it "shows a relatively weak connectedness of classes, resulting from the small number of associations and generalizations between

the classes” [5]. This reasoning is of utmost relevance to our objective of easing the development of networked game development with good API designs in Tundra. We pursue a limited set of powerful abstractions with clear interactions that a game developer could easily learn and grow to master. Not all source code lines are equal - a poorly designed API makes it a struggle to get even a few operations working if the developer has to strive for functionality scattered around in an incoherent way.

The Object-Points, as applied here, are a sum of two parts: Class Points (CP) and Message Points (MP).

**Class Points (CP)** (Eq. 1-4) are calculated from the static class structure: the class count and sums of attribute, operation and relation counts. Weights are employed to correct the values for the overall calculation. Class inheritance is taken into account by calculating novelty weights for specializing classes.

**Message Points (MP)** (Eq. 5-8) are defined by the set of operations (functions/methods) *actually used* in the software. First, the number of operations is recorded. Then, the parameter count for each called operation is collected. Also, the source and target counts of the operation calls are established. Again, novelty weights are used to compensate for repeated occurrences due to subclassing.

$$CP = \left( W_C |C| + \sum_{c \in C} |A_c| + W_{R_c} \sum_{c \in C} |R_c| + W_{O_c} \sum_{c \in C} |O_c| \right) \overline{N_C}, \quad (1)$$

where

$$\overline{N_C} = \frac{\sum_{c \in C} N_c}{|C|}, \text{ and} \quad (2)$$

$$N_c = \begin{cases} 1, & \text{if class is novel} \\ 0.5, & \text{if class has a super class} \end{cases} \quad (3)$$

$$MP = \left( W_{O_M} |O_M| + \sum_{o \in O_M} |P_o| + W_{S_o} \sum_{o \in O_M} |S_o| + W_{T_o} \sum_{o \in O_M} |T_o| \right) \overline{N_{O_M}}, \quad (4)$$

where

$$\overline{N_{O_M}} = \frac{\sum_{o \in O_M} N_o}{|O_M|}, \text{ and} \quad (5)$$

$$N_o = \begin{cases} 1, & \text{if operation is novel} \\ 0.5, & \text{if operation is provided by super class} \end{cases} \quad (6)$$

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$C,  C , \dots$	Set of classes, Class count
$ A_c , \dots$	Attribute count per class
$ O_c , \dots$	Operation count –”–
$ R_c , \dots$	Relation count –”–
$N_c, \dots$	Novelty weight of class c
$\overline{N_C}$	Avg. class novelty
$ O_M , \dots$	Set of called operations
$ O_M , \dots$	Called operation counts
$ P_o , \dots$	Parameter count of operation o
$ S_o , \dots$	Source count –”–
$ T_o , \dots$	Target count –”–
$N_o, \dots$	Novelty weight –”–

The weights are adopted directly from the earlier usage of OP for API complexity analysis, which further uses the standard Data Points analysis values by Sneed [11].

2) *Extracting Object-Point Data from Source Code*: To obtain the static class data for the Class Points (CP), we utilize existing source code parsing and annotation systems in the API documentation tools. The first alternative implementations for a minimal networked game on different modern high-level APIs studied here are written as a Javascript application and a combination of Actionscript (as3) for the client and Java for the server module. We have developed parsers for the internal intermediate representations of class and method signatures in JsDoc JSON and AsDoc XML formats. The class information is read by a Python application to an internal model which contains the data for the OP calculation, implemented in another module in the same Python application.

To calculate the Message Points reflecting the dynamic function calls, we use the Closure Javascript compiler to traverse the source code to collect function calls and their argument counts. A parser made with Python is used to read the function call data required to calculate the MPs.

While our tool calculates complete Class Point data, it currently omits two factors in Message Point data: the source and target counts of the interactions, and the novelty weight. While the tool tallies separate calls to each called function in the source code, it is not yet clear how to map them into the MP values. Sobernig et al. always set the source and target counts to 1. For the novelty weight we should check for each called operation called whether it is implemented in that class or inherited from a superclass. Our tool does not currently know the class of the object in which an operation is called. These omissions are not expected to affect the OP values significantly, at least not to the extent undermining the conclusions of this study.

Finally, to facilitate manual validation and visual communication of the data extracted from the source codes, our tool also creates UML class diagrams from the very same in-memory data structure that is used in the OP calculation. We chose the UXF format of the open source Umlet GUI diagram tool, due to its simple and straightforward XML format and the even simpler plaintext syntax used to describe individual UML elements, such as a class or a relation. This allows further manual editing of the diagrams with the GUI tool to improve the layout and annotation with notes. We are not aware of any previous implementation of directly extracting data for the OP method from the source code. Repository based automatic queries for OP analysis have been presented earlier by Henrich [14]. There, a repository of documents, or abstract software design models (PCTE), was queried using the P-OQL language.

## B. JSComplexity

JSComplexity is a recently released tool for analyzing the software complexity of JavaScript projects [6]. It reports four different types of complexity metrics of which we use the cyclomatic complexity and the maintainability index. Cyclomatic complexity introduced by McCabe [15] counts the number of linearly independent paths through a source code. Lower values are better and McCabe suggested using ten as

a threshold value, beyond which modules should be split into smaller units. The maintainability index introduced by Oman et al. [16] is calculated as follows:  $171 - (3.42 * \ln(\text{mean effort})) - (0.23 * \ln(\text{mean cyclomatic complexity})) - (16.2 * \ln(\text{mean logical LOC}))$ .

The maintainability index gets values on a logarithmic scale ranging from negative infinity up to 171, with larger values indicating a higher level of maintainability. Oman et al. identified 65 as the threshold value below which a program should be considered difficult to maintain.

#### IV. CASE SETUP

##### A. Pong as a Reference Game

We propose using the Pong game as a minimal networked multiplayer reference game in the subsequent API complexity analysis. While Pong is tiny in its functionality, it is still sufficient for demonstrating key challenges in networked games, given the functional combination of the clients controlling their own paddles and the ball bouncing in the shared space. Pong has been used in networked game research earlier, recently in an interesting study of latency compensation techniques [17]. Also, even a minimal game such as Pong reveals the amount of software needed for the basic functionality: launching the networked game, establishing connections, handling players joining in and dropping out, and synchronizing gaming events.

##### B. Game Platforms

We briefly introduce the two game platforms whose API complexity we are going to compare, the open source realXtend Tundra SDK [4] and the Union, a proprietary closed source product<sup>1</sup>. Both are relatively high-level platforms for networked games and bear several interesting similarities and differences for this study. Both are specifically designed for networking, which is exposed to the developer at an abstract application level. That is, the developer and the game do not have to know anything about sockets or network hosts. Instead, an abstract container object is provided (Room in Union, Scene in Tundra) and the game application logic listens to events from the container, for example when a new client joins the shared session/space. Also, both platforms provide an automated mechanism for synchronizing shared state over network. The shared state is stored in special attributes (objects of type Attribute) residing in the container (in Union directly in the Room object, in Tundra in the Components of the Entities in a Scene). The attributes are automatically shared among all participants, and notifications are provided for parties that have subscribed to be notified of changes. This way it is simple to for example set the game scores on the server, and show it in the client GUIs.

However, the two platforms also have fundamental differences and we discuss how they manifest in the implementation of the Pong game. TundraPong<sup>2</sup> was implemented by the leading author of this paper and an independent developer in two sessions totaling about six hours. UnionPong<sup>3</sup> was

downloaded from the Union website, where it is available as a tutorial example. While TundraPong is a script running atop the Tundra platform, UnionPong is a client application, to which networking functionality has been added by using Union's Reaktor Flash library. TundraPong utilizes a complete static scene datafile, where the game logic moves objects around. It runs on an existing client-server system, and utilizes several default components of the platform, most notably the data defining visual appearance and spatial instantiation and movement. In contrast, UnionPong not only has code to create the appearance of the game court (as it is called in Court.as), but also to define the data required for the spatial movement of an object (PongObject has x, y, direction, speed, width and height). In TundraPong, the predefined built-in Placeable component contains the position and the RigidBody component contains shape information for collisions and speed vector for movement.

Thus, it is clear from the outset that UnionPong is more complex, due to the game code containing a much larger proportion of the implementation of the functionality. The upcoming API complexity analysis is still useful as it helps to answer the questions at hand: a) how the two APIs succeed in hiding the complexity from the developer and b) how our tool succeeds in evaluating the relative complexity of the two APIs.

##### C. API Complexity Data

The complexity metrics of the two Pong implementations is presented in Table 1. We first discuss the OP values provided by our tool. As anticipated, TundraPong has clearly smaller OP values. For TundraPong the OP data is extracted from the single Javascript source file (assets/game.js) that contains both client and server functionality in two respective classes, with GUI and minimal game session management. UnionPong has separate client and server source code files in different languages using different libraries. Therefore, to facilitate more equal comparison, for TundraPong we also provide the OP data for the client only, even though it is included in the same source code file.

For UnionPong the OP data is calculated for all 14 client side ActionScript files and for selected 8 classes related to networking (GameManager, GameStates, KeyboardController, PongClient, PongObject, RoomAttributes, RoomMessages, UnionPong). The excluded classes cover GUI, the 2D scene implementation and general settings and utilities (clamp, ClientAttributes, Court, HUD, Rectangle and Settings). KeyboardController is included because it sends remote control messages from the player to the server (modifies client.paddle's attributes and says client.commit()).

UnionPong's Java server component (PongRoomModule.java) contains two classes: PongRoomModule (implements Module, Runnable) and PongObject, which is basically a duplicate of the same class in the client. As our OP data

<sup>1</sup><http://www.unionplatform.com/>

<sup>2</sup>[https://github.com/realXtend/doc/tree/master/api\\_complexity/PongMultiplayer](https://github.com/realXtend/doc/tree/master/api_complexity/PongMultiplayer)

<sup>3</sup>[http://www.unionplatform.com/?page\\_id=1229](http://www.unionplatform.com/?page_id=1229)

TABLE I  
COMPLEXITY METRICS FOR THE TWO PONG IMPLEMENTATIONS

metric	Tundra Pong		Union Pong		
	Full	Client	Client Full	Client Net	Server
LoC	361	115	565	420	281
C	2	1	14	8	2
CP	75	27	180	140	75
MP	103	63	196	175	87
OP	178	90	376	315	162
$\Sigma$ Cyclom.	35	16	58	49	-
max Cyclom.	11	11	11	11	13
Maint.Idx.	100	105	119	116	-

extraction tool does not yet support Java, we collected the OP data from the server component manually.

Overall, the OP values appear useful for characterizing the overall complexity. The values seem to reflect the functionality that the application needs to cover with its own code, versus the ready-made functionality provided by the platform — complete scene geometry, physics and basic client-server functionality in Tundra.

Regarding the values from the JSComplexity tool, the cyclomatic complexity is fine-grained, scoring points for individual functions. It is relevant as the actual programming work is eventually done in functions, whereas class structures and such are only containers for code, after all. The aggregate cyclomatic complexity value is not very useful as it is simply a sum of the values of individual functions. That is, even the simplest functions with the minimal complexity value of 1 increase the sum, thus the aggregate value tells more about the size of the codebase than its complexity. The aggregate cyclomatic complexity values correlate with Sneed’s class points (CP) as they involve also operation count. The real usefulness of the cyclomatic complexity is in identifying peak values, i.e. individual functions with high complexity. Interestingly, the maximum cyclomatic complexity value of 11 occurs in a similar functions in both reference implementations: TundraPong’s GameServer.updateGame, GameClient.onServerData and UnionPong’s GameManager.updateBall. These functions implementing the basic logic of ball movement and bouncing are indeed quite similar. This alludes that the complexity of the actual game mechanics is quite equal in both implementations. This in turn suggests that the two platforms do not differ much with respect to the complexity of basic mechanics such as moving objects despite the software implementations being quite different. The same maximum cyclomatic complexity value of 11 in the TundraPong client side function onServerData is just a coincidence: the complexity is due to the game session handling commands being made with JSON and handled in if blocks in a single function. But again, the equivalent updateBall function implementing ball movement, bouncing and boundary checks in the UnionPong server has the highest cyclomatic complexity value of 13.

Finally regarding the maintainability indices it is worthwhile to note how the values are better for the otherwise

larger and more complex Union Pong implementation. This is due to the different programming styles of the reference implementations. TundraPong has a smaller number of larger functions yielding a higher maintainability index. In contrast, UnionPong has many minimal classes and small functions resulting in a low index. TundraPong’s maintainability index of 100 is still well above the 65 threshold proposed by Oman et al. In any case, these partly contradicting complexity values highlight how the reference implementations should be as comparable as possible in terms of structure and programming style, to leave the API as the only potential source for the complexity. We propose such carefully crafted reference implementations to be made in future work.

#### D. UML Diagrams

Figures 1 and 2 show the UML diagrams generated from the OP data by our tool for subsequent manual verification of the analysis and the API complexity.

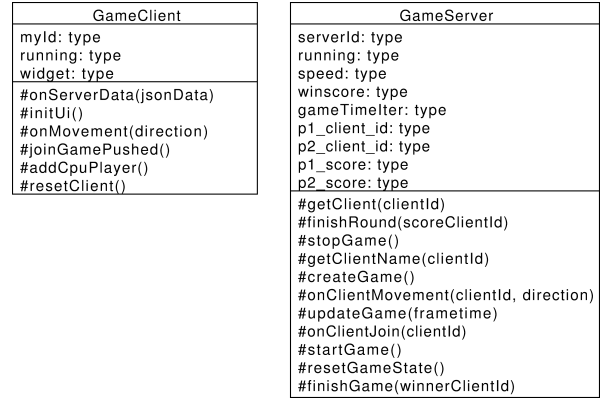


Fig. 1. The two classes in TundraPong game.js.

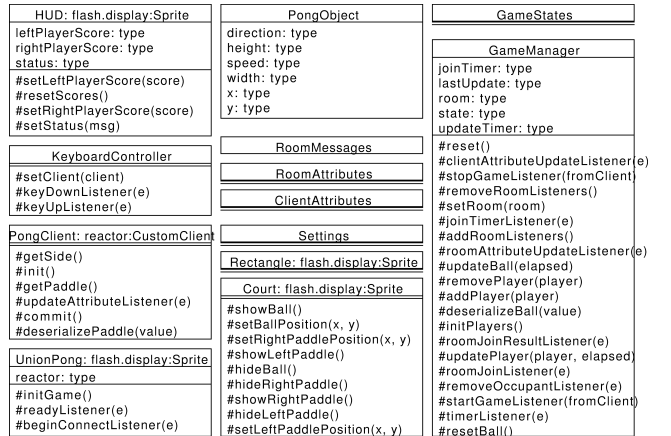


Fig. 2. The 13 classes in UnionPong client side ActionScript.

## V. DISCUSSION

### A. Interpretation and Validity of Object-Point Values

The fact that UnionPong obtains much higher OP values indicating a more complex API does not mean that the Union platform would be somehow inferior. Instead, it highlights the nature of game development at a different abstraction level. As we discussed in Section IV.B, the platforms achieve the basic functionality of the Pong game, such as synchronizing object movements and ball collisions and bounces, in different ways: UnionPong uses game specific messages, whereas TundraPong relies on the built-in functionality of the Tundra platform. Otherwise, the two APIs are very similar regarding networking. Both have an abstract container for the state, a Room in Union, and a Scene in Tundra. An application can store own custom state information as special attributes in the container, and the system takes care of automatically synchronizing changes to the state information. Both use callbacks heavily, for example to listen to new clients entering the service (an event of Room in Union's Reaktor and in the RoomModule on the Union server separately, an event of the Server core API object in Tundra server) and to attribute changes received from the network. They both also allow sending simple ad-hoc custom messages: Tundra uses them for game events such as informing of a victory with the associated data; UnionPong uses them for networking, including also paddle and ball movements, which Tundra does automatically. These similarities indicate that the OP analysis effectively captures the aforementioned differences in the scope and the abstraction level of the platforms.

Looking at the OP data and considering the OP method, our interpretation is that it succeeds in illustrating the difference in scope and abstraction level between the two codebases. We have to ask whether the OP method does that better than some other, perhaps simpler, metrics would do. From previous research we know that the OP method does succeed in identifying complexity that a simple LoC metric would miss [5]. Our data seems to support that same conclusion, as the LoC measure would give even larger complexity difference between the two implementations (115:565 for full clients). Based on qualitative analysis, we think that the smaller relative difference indicated by the OP data is more appropriate. Even though UnionPong client needs to do more, and especially has many more classes, most of the classes are very simple and most of the source code is not very complex. Considering only static class information, the difference would be even greater (27:180). TundraPong has relatively long methods and a lot of function calls, which lead to relative high MP count in the dynamic analysis (63:196). We think that changes the final OP score to a realistic ratio (90:376). Based on the OP data we cannot really say whether the tool and the OP method miss something essential in the API complexity analysis. For example, the OP method does not take into account anything specific to networking: the need to think of connections, defining and sending network messages etc. They are of course accounted for as normal data definitions and function calls, but

would some networking specific metric, for example for the number of messages, be more useful instead? Arguably, they present an additional complexity that the game developer has to manage.

### B. Built-in Platform Logic vs. Custom Application Logic

As TundraPong shows, implementing a game on an rich platform such as Tundra can require a comparatively small amount of work. However, custom game specific solutions for game object data, network messages, movement inter/extrapolation and collisions can easily be more powerful and even required. For example, if the game takes place on a small spherical world, a mini planet, Tundra's built-in Euclidean movement techniques become suddenly much less useful. Therefore, the logic underlying the Union platform and other similar smaller libraries is sound. Game developers often need custom solutions, so the platform just provides the lower level tools for messaging and stays out of the way for the rest. However, optimizing and perfecting for example movement synchronization is not trivial, thus it is often useful to have a mature shared implementation for common operations.

### C. Complexity Metrics in Estimating Ease of Development

Hatton criticized that software complexity metrics, including cyclomatic complexity, have poor correlation with actual reported defects in large software projects [18]. He argued that none of the various complexity metrics gives any better prediction than the simplest metric of lines of code. In our study, however, the objective is not to predict defects but to estimate development effort and ease of development using particular APIs.

The recent metrics for the cognitive complexity of software may thus be suitable, for example the Cognitive Information Complexity Measure (CICM) reportedly satisfies all the criteria set for an effective software complexity measure [19]. It draws from research in cognitive informatics and is based on calculating basic control structures (BCS) from the code and assigning cognitive weights for those constructs. The studies on cognitive complexity focus on comprehensibility of the software when the source code is read. We assume that it correlates with the ease of developing that piece of code as well. However, we have been unable to find any automated tools to calculate CICM or other BCS based metrics. Also, their objective is to analyze complexity within individual functions, similarly to cyclomatic complexity, whereas in API evaluation the overall program structure may be more relevant.

### D. Limitations and Future Work

The Pong game used as the minimal reference of a networked multiplayer game in this study is very simplistic. Much of the complexity of real networked games, and especially large scale commercial MMOGs, lies in areas not addressed by this study: service reliability, availability, restorability and scalability [1]. Networked programming in general is also typically complex due to the need to handle several kinds of error situations, such as lost data, dropped connections

and conflicts from simultaneous actions. The Pong reference implementations of this study may well be limited in that they do not handle such issues in the way a production quality game must handle, which probably increases the complexities. However, both Pong games are built on very high-level networked game platforms, which strive to hide the complexity of networking from the developer. Whether and how they really achieve that cannot be determined from the data of this study, but would require a different analysis.

In future work the shortcoming of relying on too simple reference games in API complexity analysis could be addressed in two ways. First, we could analyze codebases of real production games. However, typically a particular game exists only as a single implementation, which prevents comparative analysis. Nevertheless, the analysis could still provide valuable insight into assessing the evolution of the complexity of the game, and the correlation of the complexity with the monetary expenses of the development effort. Second, we could develop a more complex reference game and ensure its completeness. Such a reference game should be carefully specified to cover all relevant areas of networked gaming, but still remain small enough to allow implementation within a realistic timeframe. Existing canonical implementations may provide a starting point, as for example several commercial networked 3D first-person shooter (FPS) games have been open sourced (Quake, Cube2), and at least one high-level platform already features a FPS as a tutorial (Torque3D).

We propose a collaborative effort, similar to the TodoMVC.com for web platform comparisons [12], to work on reference games for API evaluation purposes. We suggest using *refgames.org* for the purpose and provide these example Pong's and our OP tool there as seed material.

## VI. CONCLUSIONS

Our study shows that software complexity metrics can be useful aids in comparing APIs but so that human judgment is exercised in assessing the plain numerical complexity values. The different metrics give useful information on the overall characteristics of code bases and can also highlight potential complexity hotspots for further study. A collaboratively composed collection of reference games on various platforms seems like a fruitful way to get rich API comparisons in the future.

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