

Characterizing User Behaviours and Products in an Online Educational Community: A Comparison between Novices and Elders

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ABSTRACT: This study focuses on two user types in an online educational community — novices and elders — and explores the characteristics of their behaviours and contributed products by analyzing usage logs. Results show that novices and elders have different behavioural trajectories. In addition, the products created by elders attract more views.

KEYWORDS: Online educational community, novices, user behaviours

1 INTRODUCTION

Many online educational communities (OECs), such as the *Instructional Architect* and *Tapped In*, have been established to facilitate user knowledge building, promote mutual interaction, and meet educational needs (Preece & Maloney-Krichmar, 2003; Recker, 2006; Suthers & Dwyer, 2014). By examining time spent in an OEC and the levels of participation, different user types can be identified. For example, we define *novices* as new members of an OEC, and *elders* as users who have participated in an OEC for some time and make frequent contributions (Bishop, 2007). Investigating the differences between novices and elders is important, as it allows researchers to understand the evolution of user behaviours and characteristics of the products they contribute (e.g., discussions, comments, educational artifacts). It also offers implications on the sustainability of OECs.

The purpose of this study is to identify the differences between novices and elders based on the usage logs from a free, web-based tool called the *Instructional Architect* (IA.usu.edu). The IA allows teachers to build shared repositories of educational content and support each other. Teachers can contribute to repositories by *creating* new products, *modifying* existing products, and *adding external resources* to their products (Recker, 2006). They can support each other by publishing their products, so other users can *view* and even *copy* their products. In particular, this study compares the behaviours and resulting products between 386 *novices* who created accounts in the IA during 2013 and 152 *elders* who created accounts before 2013 and actively participated during 2013.

2 THEORETICAL CONTEXT

The rapid growth in Internet technologies has given rise to many OECs, which provide features to meet

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teachers’ instructional needs (Andrews, Preece, & Turoff, 2001). The way users use features of an OEC varies. For example, some prefer to create products using their own knowledge, some tend to build on and remix external resources in their products, while others may simply view other users’ products (Yuan & Recker, 2014). In addition, user behaviours in an OEC are dynamic across time. For example, novices may be initially attracted to an OEC, and some eventually become key contributors and elders, while others leave after a short time (Iriberry & Leroy, 2009). In this way, the dynamics of user behaviours can affect the sustainability of the whole community.

Further, Farzan, DiMicco, & Brownholtz (2009) noted that users’ products can affect the sustainability of an OEC, as these products can help attract new users and influence existing users to remain in the community. Thus, describing user behaviours and characterizing users’ products can lead to a better understanding of the sustainability of an OEC. However, little research has focused on both aspects, especially drawing upon the usage logs of a specific community. In response, this study compared behaviours and products between novices and elders, and concludes with suggestions on how to improve the sustainability of OECs.

3 METHOD

This study addresses two questions oriented towards the IA community: 1) How did novices differ from elders in terms of their behaviours in 2013? 2) How did novices differ from elders in terms of the products each group created during 2013? We collected data from the IA database and analyzed it using descriptive statistics. We also used the Mann-Whitney test due to non-normal distribution.

4 RESULTS

4.1 A Comparison of User Behaviours

First, on average, novices created more products, but made fewer modifications and incorporated fewer external resources than elders (see Table 1). However, the Mann-Whitney test reveals that these differences are not significant. Second, novices and elders show different behavioural trajectories. In general, novices exhibited high activities after creating their accounts, after which their activities decreased. For example, novices created almost 50% of their products during the month they joined the community. On the contrary, elders made steady contributions to the IA throughout 2013.

Table 1: Comparison of behaviours between novices and elders.

Metrics	Novices (N=386)			Elders (N=152)		
	Mean	Median	SD	Mean	Median	SD
# of products created	1.32	0	2.50	1.16	0	2.43
# of modifications of products	22.38	0	45.37	23.51	0	53.75
# of external resources added to products	3.62	0	11.08	5.65	0	14.60

4.2 A Comparison of Resulting Products

During 2013, novices created 510 products and elders created 177 products, but elders’ products were viewed significantly more times ($U = 31637.50$, $p < .01$; see Table 2). This suggests that elders contributed products that are more beneficial to users.

Table 2: Comparison of products between novices and elders.

Metrics	Novices’ products (N=510)			Elders’ products (N=177)		
	Mean	Median	SD	Mean	Median	SD
# of times products being viewed*	12.41	3	43.25	133.32	17	388.04
# of times products being copied	.01	0	.11	.03	0	.20

* Difference between the two groups is significant (Mann-Whitney test; $p < .01$)

5 CONCLUSION

This study offers a series of suggestions on improving the sustainability of OECs, such as providing support and dissemination to retain novice interest, especially within the first few months after they join the community, and highlighting elders’ products to attract more new users. Furthermore, this study contributes to learning analytics, as it offers examples of how to trace and compare user behaviour and users’ products based on usage logs (Ferguson & Buckingham Shum, 2012). This work is part of an ongoing series of studies; a forthcoming study uses logistic regression to investigate which variable(s) predict whether novices remain in the IA community.

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