

User-Centered Design for Knowledge Imbalance Analysis: A Case Study of ProWD

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Abstract. Not all topics within a knowledge graph are represented at the same depth, which might lead to biased conclusions derived from the graph. Tools have been created in an effort to highlight knowledge imbalances, one of which is ProWD, built to analyze imbalances in the Wikidata knowledge graph. However, as often found, the usability aspect of Semantic Web tools is commonly overlooked, resulting in their limited acceptance. In this paper, we utilize the standard approach to improve usability, i.e., the User-Centered Design (UCD), for ProWD. We employ the full range of steps of UCD to improve the user experience of ProWD despite the complex nature of knowledge graph concepts underlying ProWD. The result of the ProWD redesign is then evaluated using the System Usability Scale (SUS) and User Experience Questionnaire (UEQ) scores, as well as the task success rate and completion time, suggesting that the overall usability of ProWD has successfully improved.

Keywords: Knowledge Imbalance · Wikidata · User-Centered Design

1 Introduction

The increasing utilization of knowledge graphs (KGs) makes it necessary to ensure their quality. One of the most popular KGs is Wikidata, which is part of the free-content Wikimedia family with its main goal to collaboratively collect structured data to be used by anyone [15]. In a commercial context, Wikidata is used by companies like Amazon⁵ and Google.⁶ In terms of quantity, there are over 1.1 billion statements describing 88 million items⁷ on Wikidata about various topics, ranging from humans and cats to movies and museums. Given such a vast quantity, the question arises as to whether all topics within Wikidata (or any KG, in general) are represented in a well-balanced manner. This question is particularly important as failing to notice the existence of imbalances in a KG might lead to misleading conclusions derived from the KG.

⁵ <https://www.wired.com/story/inside-the-alex-friently-world-of-wikidata/>

⁶ <https://ahrefs.com/blog/google-knowledge-graph/>

⁷ <https://wikidata-todo.toolforge.org/stats.php>

Related Work. In terms of data quality, the problem of analyzing the existence of knowledge imbalances touches mainly two aspects: completeness and coverage. Data completeness concerns the degree to which all necessary information is provided, whereas data coverage refers to the level of detail of the information stored [17]. Knowledge imbalances occur whenever there are disparities of completeness as well as coverage between different topics in a KG.

In the context of Wikidata, identifying and measuring knowledge imbalances are deemed to be an important step to advance knowledge equity in Wikidata [18]. A number of initiatives have been undertaken to address knowledge imbalances in Wikidata. Denezh is a tool developed to track the gender gap in Wikidata [7]. Via Denezh, one may, for example, compare the number of male vs. female humans in Wikidata. In [1], a map-based visualization shows which locations worldwide have how many Wikidata items. Through the visualization, one might observe that there is a stark difference between the number of Wikidata items located in North America vs. South America. Recoin [2] measures the different levels of completeness among Wikidata items, ranging from very basic information to very detailed information.

ProWD. Let us now draw attention to ProWD. ProWD is a framework to measure knowledge imbalances in Wikidata [16]. ProWD generalizes the above initiatives in the sense that ProWD can be leveraged to any domain instead of only the gender or geographical domain. The framework measures knowledge imbalances based on Class-Facet-Attribute (CFA) profiles. A class groups similar items, which might comprise multiple facets, thus allowing attribute completeness to be compared. For example, via ProWD one may compare the completeness of the attribute “date of birth” and “educated at” between US inventors vs. Nigerian inventors, as illustrated in Fig. 1. In the figure, we observe that 95.65% and 49.07% US inventors have the “date of birth” and “educated at” information, respectively. On the other hand, 100% (i.e., all) and 83.33% Nigerian inventors have the “date of birth” and “educated at” information, respectively. Moreover, the absolute number of Nigerian inventors seems to be far less than that of US inventors. The ProWD tool is available at <http://prowd.id:3333/>.⁸

Problem and Contribution. While ProWD indeed offers basic functionalities in identifying knowledge imbalances in Wikidata, little concern was given to the extent to which the ProWD tool can be used by its users to achieve their goals with effectiveness and satisfaction. This very notion of usability [10] is nevertheless commonly overlooked by KG application developers [4]. The novelty and intricacies of the KG concepts underlying ProWD as well as the fact that ProWD was developed without any rigorous user analysis and testing, raise questions about its usability.

In an effort to improve the usability of ProWD, one could prioritize the (potential) users in such a way that their particular needs can be accommodated more accurately. A process commonly used in the development of a product, putting the user experience and usability front and center, is the User-Centered Design (UCD). The UCD approach is widely investigated and utilized in academia and industry [12]. Adapting and implementing the UCD approach towards the development of ProWD is challenging for two

⁸ A demo video of the tool is also available at <https://youtu.be/3jcXXx1uQU4>.

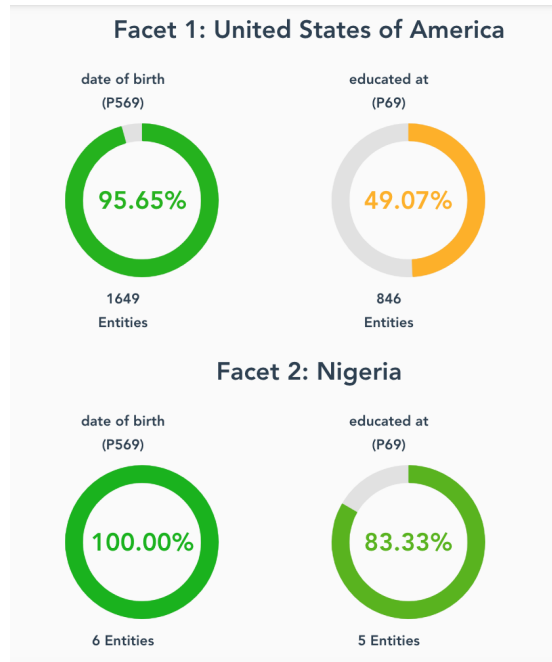


Fig. 1. Comparison of the completeness of attributes between US inventors and Nigerian inventors on the original ProWD (Source: <http://prowd.id:3333/#/profile/compare/123>)

reasons: (i) knowledge graphs are a new abstract concept, and (ii) the imbalance analysis of such a graph is even more abstract. With that being said, this paper contributes to the adaptation and implementation of the UCD approach in the KG setting to improve ProWD, a knowledge imbalance analysis tool.

Paper Structure. The rest of the paper is outlined as follows. Sec. 2 provides a brief background of the research. The next four sections correspond to adapting the steps of UCD in improving the usability of ProWD: understanding ProWD's context of use (Sec. 3), specifying ProWD user requirements (Sec. 4), redesigning ProWD (Sec. 5), and evaluating the redesigned ProWD (Sec. 6). Sec. 7 concludes our paper.

2 Background

Knowledge Graphs. A knowledge graph (KG) describes real-world entities and their relationships [3]. Resource Description Framework (RDF) is the standard data model by W3C for KGs [8]. For example, the statement Bob is a human can be modeled in RDF as a Subject-Predicate-Object triple (Bob, is, human). To query RDF data, one could use SPARQL, which provides rich query constructs, enabling the creation of complex SPARQL queries [8]. Wikidata is an open, cross-domain KG, providing structured data to anyone [15]. Wikidata provides RDF support which can be queried via its SPARQL endpoint.

User-Centered Design. UCD is a design process comprising four activities: *(i)* understand the context of use, *(ii)* specify user requirements, *(iii)* design alternative solutions, and *(iv)* evaluate against requirements [10]. UCD prioritizes user needs in the design and development process of a product. To evaluate a product in UCD, several usability metrics can be used, such as the success rate and time on task, as well as questionnaire-based metrics like the System Usability Scale (SUS) [14] and User Experience Questionnaire (UEQ) [13]. The SUS measures the overall usability of a system, while the UEQ measures six scales of a product: *(i)* attractiveness, *(ii)* perspicuity, *(iii)* efficiency, *(iv)* dependability, *(v)* stimulation, and *(vi)* novelty.

3 Understanding ProWD’s Context of Use

To improve upon an existing application, one must understand the context of use of the application. From this section onwards, we refer to the original version of ProWD [16] as `ProWD-V1` and the redesigned version of ProWD resulting from our research as `ProWD-V2`. ProWD was developed in a technology-driven manner and was made to shed light on imbalances that might exist within Wikidata. This technology-driven development led to the application being less user-friendly as the users were not involved in the development process.

The creators of ProWD wishes to advance ProWD to be utilized more widely by users outside the Wikimedia community (WM Community). The creators observed and concluded that the main challenges of ProWD are: *(i)* the abstract notion of the Wikidata knowledge graph and its imbalance analysis, and *(ii)* the lack of an existing ProWD user base. To tackle those challenges, one needs to be able to convey properly such abstract concepts, and to analyze (and accommodate) the generic and specific user requirements of ProWD.

To confirm this, we initiated a meeting with representatives of the WM Community of Jakarta (the capital of Indonesia) to gather feedback and insights about their expectations towards ProWD. The WM Community of Jakarta is considered to be representative of the Wikidata Community as they are experienced with Wikidata and are interested in the issue of imbalance and the potential of ProWD as a tool. The meeting uncovered several potential usability issues. The target users identified were as follows: *(i)* data journalists, as Wikidata could become a data source for news topics, *(ii)* data professionals, as Wikidata might enrich their data analysis, and *(iii)* AI researchers, as Wikidata could be utilized for training data to build AI models.

4 Specifying ProWD User Requirements

Having identified the potential user base, we wanted to collect their requirements. To gather those requirements, the targeted users participated in usability testing and are interviewed. As `ProWD-V1` has already existed, a heuristic evaluation is done to gather the requirements of how the application can be improved [6]. Hence the process of specifying the requirements is two-fold: *(i)* heuristic evaluation to gather generic user requirements, and *(ii)* evaluation based on specific requirements from the appropriately chosen test-participants from the defined user base.

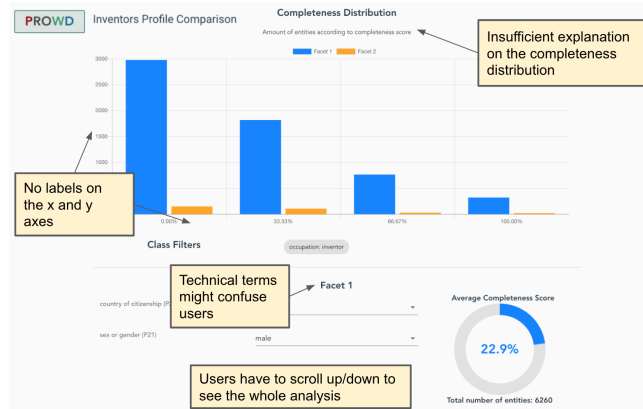


Fig. 2. Compare two facet values feature on ProWD-V1

4.1 Heuristic Evaluation

The heuristic evaluation using Nielsen’s ten usability heuristics [11] shows how the use of technical terms with a lack of further documentation in ProWD-V1, hence confusing the users. An example is the term “Class-Facet-Attribute” in the landing page of ProWD-V1, which confuses the users as there is no further information on what it exactly means and how it ties to the purpose of the application.

Within the “see full profile”, “compare two facet values”, and “multi-dimensional analysis”, the users are burdened with a large cognitive load as they need to keep scrolling back and forth to look back within the same page, which is against the “Recognition rather than recall” heuristic. ProWD-V1 is lacking in providing alerts about the status of the system, for example, whether a page is loading or an error occurs for the page. Fig. 2 illustrates such usability issues in ProWD-V1.

4.2 User Research and Testing

Selection of Test-Participants. The test-participants were chosen based on the context of use, namely, WM community members and potential Wikidata users. A total of five users from each distinct user group were selected. In selecting the test-participants, the method used was purposive sampling.⁹ To lessen the bias as much as possible, we chose test-participants who would be critical towards systems such as ProWD based on their experience, occupation, and their sensitivity and interest towards knowledge, culture, data, and education.

The selection process led us to have five WM community members and five potential Wikidata users comprising data analysts, journalists, and teachers/researchers: (i) the WM community members are all from Jakarta and are those in high-profile positions which presumably would have valuable and critical feedback for ProWD as they have the biggest drive for ProWD’s purpose; (ii) since ProWD’s functionality is quite

⁹ That is, intentional selection of informants based on their ability to elucidate a specific theme, concept, or phenomenon.

similar to that of a dashboard, data analysts would be familiar with the kind of task that ProWD supports and would be able to give valuable feedback; *(iii)* lastly, the chosen journalist and teachers are people who are highly sensitive towards social/cultural issues, e.g., gender bias.

User Research and Persona. To create a clear and constant reference to the users, a persona can be created [9]. The test-participants were interviewed to discover more about their personality, needs, and expectations using a semi-structured interview. The interview reveals that 100% of the test-participants have previously encountered a form of imbalance in general knowledge, while 90% state that more information on the imbalances would be insightful for their work. These confirm that ProWD has a potential in helping these users. From the interview, we also observe that users have a varying degree of Wikidata proficiency, so we divide the target users into two main personas. The first persona is the *WM Community Member*, who is an expert on Wikidata structures and is tenacious in learning new systems. They have a drive for general knowledge and want to highlight underrepresented topics. The second persona is the *Knowledge Enthusiast*, who has no experience with Wikidata, though is potentially interested with it. The knowledge enthusiasts value reliability and are sensitive towards their domain of interest. Each of the two personas has their own end goal: The *WM Community Member* wants to understand the knowledge imbalance situation to be able to prioritize her actions, while the *Knowledge Enthusiast* wants to have an understanding of the quality of general knowledge and identify topic imbalances.

Usability Testing Setup. Now that we have gathered information regarding the users, we want to know how users complete tasks using `ProWD-V1`. From observing the users in completing tasks, we could gather requirements on the things to keep and things to improve from `ProWD-V1`, for this purpose, the users participated in a usability testing session using six tasks formulated based on the heuristic evaluation and context of use of ProWD. From the usability testing, specific requirements in the form of qualitative feedback and quantitative data in the form of success rates, time on task, SUS, and UEQ were collected. The collected data was then analyzed to measure the impacts of the redesign (`ProWD-V2`). The six tasks were categorized into typical and secondary tasks. The typical tasks were: *(i)* primary tasks, which are tasks to create new dashboard (CN), gather info on profile page (PP), compare subclasses (CP), discover insights (AP), and *(ii)* secondary tasks, for which the users are prompted to generate feedback on the landing page (LP), and the process of opening previously created dashboards (OP), which are the generic experience of new and returning users. The selected topic for the testing is “humans with the inventor occupation.” This topic was selected as inventors are generally known to the public, e.g., Albert Einstein. In the testing sessions, we observed how the users conducted the tasks and instructed them to think aloud to identify any cognitive obstacles throughout the testing sessions.

Post-test and Documentation. For analysis purposes, the screen and audio of the testing session were recorded. The user interview took 30-60 minutes each and required a similar amount of time to evaluate and analyze. After the usability testing session, the users answered a usability survey which consists of SUS and UEQ. Gathered from the

testing, the *qualitative feedback* suggests that even with their pre-existing knowledge, *WM Community Members* are still confused about the profile creation section of the application. The terms used on ProWD-V1 are still rather confusing and figuring out the facets and attributes of the profile becomes more difficult. Meanwhile, the *Knowledge Enthusiasts* were confused from the get-go, they neither had an idea of what the data shown means and how it is important or relevant to them, nor did they comprehend the visualizations the application provides. To simplify, coming into ProWD-V1, the users are faced with foreign terms, coupled with confusing navigation and lack of documentation, the users get easily frustrated.

5 Redesigning ProWD

Now that we have gathered the context of use and user requirements of ProWD, alternative design solutions were formulated. At this stage, several design tools and methods were utilized in the process of redesigning the application, from creating a user journey to further analyze where ProWD can become helpful for the users to an in-depth analysis of the 6 tasks. The user requirements also gave insights in the information the users want and need from the application. Hence an information architecture was created to map these information so that the users can find them more intuitively. Then a prototype of ProWD-V2 was created based on the analysis of the user requirements using the tools previously mentioned.

5.1 User Journey Mapping

As previously mentioned, each persona has their own end goal in doing and completing their job. Understanding how ProWD could help these personas reach their goals helped us orient the application towards the users' needs. For this purpose, a customer journey map was created. "A journey map is a visualization of the process that a person goes through in order to accomplish a goal." Journey maps can help determine the users' frustration, pain points, and delight. [5] The end goals of the personas which were identified from the interviews highlighted possible opportunities in which ProWD can become a helpful tool for these personas. For the *WM Community Members*, their end goal is to "be able to understand the knowledge imbalance situation better." ProWD can be helpful to this persona for comparing projects so that they can focus their efforts better by also creating an easily identifiable and comparable metric. As for the *Knowledge Enthusiasts* whose end goal is to "be able to have an understanding of the quality of general knowledge and identify topic reliability," ProWD can be helpful to identify dominating or popular topics in general knowledge.

5.2 Task Analysis and Information Architecture of ProWD

To analyze the efficiency of the 6 tasks previously tested, a task analysis is done to break-down the tasks to a finer scale. As ProWD visualizes knowledge imbalance, the tasks are partially cognitive and physical which creates a challenge in doing task analysis. To tackle this, a hierarchical analysis of physical subtasks was performed so that the users' physical interaction with the application could be broken down into its elements and analyzed to minimize any unnecessary flow. One example of a task with

both physical and cognitive challenges is the Compare Page (CP task) shown in Fig. 3. An example of minimization of a subtask is done to the “Select or Edit the Comparison Dimension” subtask, which in ProWD-V1 required the user to do tedious actions of around 11 steps to simply edit the comparison dimensions. The task analysis simplified the process by decreasing the action to just 5 steps in ProWD-V2, this optimization is done to each physical subtask.

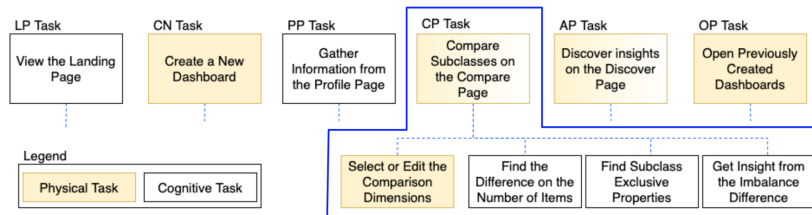


Fig. 3. Task analysis for the Compare Page (CP) Task

For the more cognitive tasks, we analyze things that can be improved based on the user feedback from the usability testing sessions. The user journey map (Sec. 5.1) shows that ProWD can help the *WM Community Member* persona by “creating an easily identifiable and comparable metric,” the creation of this metric can also be helpful to tackle the cognitive task to “Get insight from the imbalance difference” as seen in Fig. 3. Hence, the Gini coefficient was selected as a metric for imbalance. The Gini coefficient is a rough measure of the amount of imbalances in wealth distribution with a value of 0 (ideally balanced) to 1 (totally imbalanced), which is visualized by using the so-called Lorenz curve. It has been used as a measure of inequality in the economic field by calculating the Gini coefficient for income distribution of each country around the world to measure the wealth inequality. This research utilizes the Gini coefficient to visualize the imbalance in information wealth.¹⁰ Other than the mentioned examples, the requirements previously gathered also shows the need to provide additional information, e.g., “*what does ProWD enable?*” and “*how do I navigate ProWD?*”. Now that we have a plan on the information we want to add and dismiss in the form of information and features for ProWD-V2, an information architecture blueprint was created. An example of this information architecture activity is the mapping of the content within the compare page, which now includes the Gini coefficient comparison, the shared and unique properties of each subclass, additional documentation for each data, this mapping process is done for each feature within ProWD-V2.

5.3 Prototype of ProWD V2

Now that each feature’s plan is mapped, it is time to actualize these improvements into the form of an actual application by creating a prototype. The improvements done are of two kinds: (1) changes done based on the more generic UI/UX elements categorization,

¹⁰ https://en.wikipedia.org/wiki/Gini_coefficient

and (2) a more specific categorization based on tasks exclusive to ProWD’s system. The UI/UX elements improved aspects are those of (1a) *the types of information presented*, which in the new design, an addition of a new visualization of topic imbalance using the Lorenz curve to represent the Gini coefficient as mentioned in Sec. 2. A new function to be able to view properties exclusive to compared subclasses using set operations is also incorporated in ProWD-V2. Based on the (1b) *data visualization element*, to aid the users in comparing data visually, bar charts are used to represent the information over using donut charts previously used in ProWD-V1. (1c) *the flow of interaction* is also modified. The modification of the flow is the result of the task analysis previously done, an example of this is within ProWD-V2, after the dashboard creation process, the users are immediately redirected into their new dashboard instead of having to browse for it in ProWD-V1.

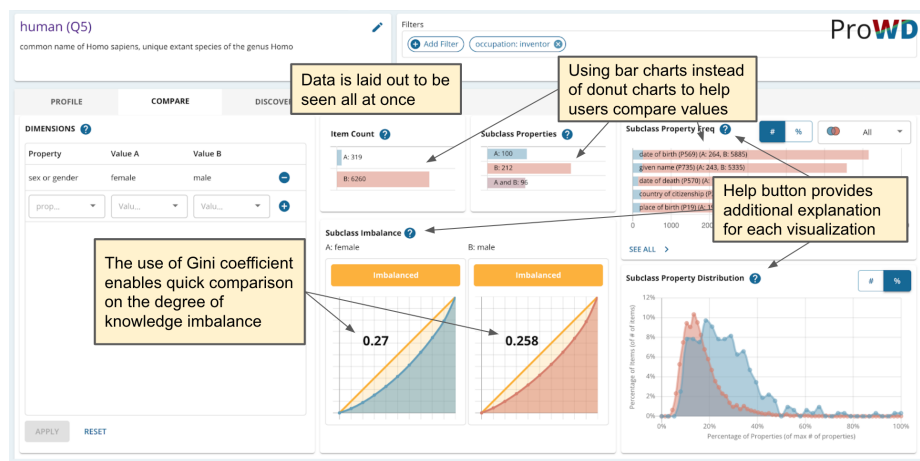


Fig. 4. Compare page of ProWD-V2

In regard to (1d) *the layouting of components*, modifications in ProWD-V2 allow the users to do less scrolling as all information can be seen in one page. Another important element is (1e) *the words or vocabulary* used to deliver the information. An example of an improvement in this regard is renaming “Profiles” as “Dashboards”, since according to the feedback from the test-participants, the term “profile” is commonly understood as a user profile, while the behavior of the feature is more similar to that of a dashboard. Instead of the “Class-Facet-Attribute configuration”, we used the term “topic” throughout the application, as this term topic again more commonly understood.

This would also tie-in with the function of ProWD, which is to visualize topics of interest. Other modification of vocabulary are renaming “Multi-Dimensional Analysis” as “Discover” and the “Gini Coefficient” as “Topic Imbalance”. Last but not least, (1f) the colors were changed in ProWD-V2. A colour palette was created as a guideline for the design of ProWD-V2, this helps to create a internal consistency for components and

actions within ProWD. A predominantly plain white background is also used to create a minimalistic interface which helps to highlight the presented data.

The more task-specific improvements of the application concern (2a) *the landing page*. By presenting an easy and simple tagline, the users are presented with a simple definition of the function of ProWD. Another characteristic within the landing page is an onboarding feature to help the users understand the concepts and background of Wikidata and ProWD. Generally, the idea of the changes on the landing page is to provide information the users need to understand and use ProWD's features. Regarding modifications more specific to the (2b) *dashboard creation task*, users now are only required to input the "class" and "filters" instead of having to select specific attributes and naming the dashboard immediately. Examples are also provided to assist the users in understanding the dashboard creation process.

Stepping into the main features of the application, (2c) within the *profile feature page*, a component was added to specifically give more information to the users when they need it, with other modifications pointing into the previously elaborated UI/UX elements improvements. The (2d) *compare page* was modified to create a more efficient flow when the users want to further specify the dimensions of the comparison they want to make, while (2e) *the multi-dimensional analysis page*, which is called the "discover page" on ProWD-V2, was altered to visualize the information in a less textual form. An example of our improvement results, can be seen in Fig. 4 which is for the compare page.

5.4 ProWD V2 Implementation

The prototype was then implemented using ReactJS¹¹ and Flask.¹² It can be accessed on <https://prowd.id>.¹³ The flow of data begins from the front-end of the application sending a request to the back-end which then fetches the live data from Wikidata query.¹⁴ By using live data provided by Wikidata's endpoint, we get the benefit of having directly updated data of the items in Wikidata. However, fetching live data also has its limitation in that it limits us to only be able to fetch 10,000 items at once.

When an analyzed topic consists of more than 10,000 items, the system will notify the user that the displayed data is in fact only a sample of the population. These 10,000 items are selected based on the Wikidata SPARQL query service's default indexing. The properties of the item, not including the external identifiers, will be considered as the wealth of each item. By assuming the properties to represent the (knowledge) wealth of each item, we can measure the imbalances of the knowledge provided by those items by comparing the number of distinct properties each item has for a certain class.

6 Evaluating the Redesigned ProWD

After the redesigned ProWD (ProWD-V2) was implemented, measuring the effects of the improvement effort could give us an insight into how UCD affected the usability

¹¹ <https://reactjs.org/>

¹² <https://flask.palletsprojects.com/en/1.1.x/>

¹³ A demo video of ProWD-V2 is available: <https://bit.ly/prowd-v2-demo>

¹⁴ <https://query.wikidata.org/>

of the application. To do so, we used several quantitative metrics which were measured in both usability testing sessions. As previously mentioned, the users fill in the SUS and UEQ after the usability testing session, while the success rate and time-on-task is gathered from the session recordings. To be able to compare the performance of the two versions, the second usability testing used the same six tasks tested during the first testing sessions, with slight adjustments to fit the changes in the flow of ProWD-V2.

An example of the adjustment is to test the Open Previous task (OP Task) last instead of second, as in ProWD-V2, the users do not need to search for the new dashboard after the creation process. The participants of this second usability testing were the same as those in the first session, except that one user was unable to participate in the second session due to health-related issues. In this case, a new test-participant, represented by the same persona as the dropped-out user, was appointed. Testing with the same users might raise a concern regarding the bias which may be brought by the users' familiarity with the application. Nevertheless, as the system has changed substantially from ProWD-V1 to ProWD-V2, the effect of application familiarity would not affect the test results severely.

6.1 Quantitative Metrics and Analysis

The SUS, UEQ, success rate, and time-on-task were compared and analyzed. The SUS results show an increase of 27 points for the two personas, promoting ProWD from being a grade F (39.75) to a grade C (66.75) application based on the SUS benchmark as mentioned in Sec. 2. This score of 66.75 suggests that ProWD's usability is barely below average on the SUS benchmark. The SUS score for the *Knowledge Enthusiast* persona is lower than that of the *WM Community Member* with the scores of 59.5 and 74 respectively. The SUS results also show that most of the users strongly agreed on the question stating that "the system requires them to learn many things prior to using the application".

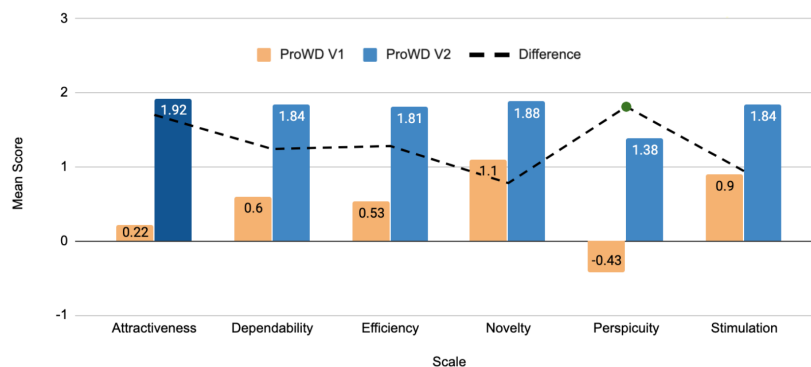


Fig. 5. UEQ Results for Each Scale

As seen in Fig. 5, the UEQ scores show an increase on each of the measured scales with the highest increase on the perspicuity scale. Though the perspicuity of ProWD-V2 is better than the one of ProWD-V1, it is still the lowest being *above average* as opposed to good and excellent for the other scales. The scales with the highest scores are novelty and attractiveness. The low perspicuity score is also aligned with the strongly agreed SUS statement previously discussed. This suggests that the users think that the application is more attractive and easier to digest compared to the previous version, though further research might be needed to make things clearer for the users.

The success rate increase of 25.9% implies that on ProWD-V2, users are more likely to successfully complete tasks. There are only 2 failed task runs on ProWD-V2 as compared to 13 failed task runs on ProWD-V1. The specific task with the highest increase in success rate is the Create New (CN) task where the users create a new dashboard, though it is the task with the lowest success rate compared to other tasks on ProWD-V2. This may suggest that the biggest hurdle in utilizing ProWD-V2 is where the users are initializing a dashboard, which may also correspond with the low perspicuity based on the UEQ. The average task completion time shows that users take the longest time on the Profile Page task (PP) on ProWD-V2. This suggests that the users might need more time to complete each subtask within the Profile Page as compared to the time needed in the same page on ProWD-V1. Though with the increase in time on PP Task, the success rate of the same task is higher on ProWD-V2 compared to ProWD-V1.

6.2 Qualitative Feedback

Generally, the usability has improved according to the quantitative metrics. Though to gather more task and user-specific feedback, the qualitative feedback needs to be taken into account. The qualitative feedback of ProWD-V2 indicates a generally positive reception of the new design. The improvements done on the *types of information presented* aspect are the strongest changes which affected the feedback positively. The users stated, “This page is nice, easy to understand, (at first glance) it looks scientific though is actually simple to use.” The new visualizations and added features on ProWD-V2 resulted in positive feedback from the users. The improvement which seems to be the weakest in effect is the *layouting/placement* aspect where the test-participants’ sentiments toward the changes does not differ greatly.

Other than the commentary feedback, the users’ behaviour throughout the testing session was taken into account. When creating a new dashboard, users tended to immediately input instead of exploring the examples. This behavior caused the users to misunderstand the necessary inputs to create a correct dashboard. Within the dashboard, the tabular information is still misunderstood by the users. This causes them to take a longer time to assess the other information in the dashboard. This feedback shows that there is room for improvement, pointing to the iterative nature of UCD.

7 Conclusions

Systems regarding abstract concepts, such as knowledge graphs and their imbalances, can benefit from adapting UCD to improve their usability. With ProWD not having an

existing user base, we had to select appropriate test-participants who are critical towards such a system and fit with the context of use. The redesign process considered UI/UX elements and task-specific improvements, resulting in a significant increase wrt. the SUS and UEQ metrics. Moreover, users are more likely to complete tasks successfully by 25.9% on `ProWD-V2` compared to `ProWD-V1`. For future work, we plan to conduct more iterations of UCD on ProWD, expand the 10,000 items limit, and streamline ProWD to regular activities by the Wikimedia community.

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