



PERSONAS FOR POLICY-MAKING AND HEALTHCARE DESIGN

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Abstract

The tool Personas has been used by different research teams to represent diversity among elderly people. This tool has been widely used in the discipline of product and service design and in recent years has spread to other fields such as strategic design and policy-making. However, these Personas lack accuracy when describing capabilities and are too general when describing psycho-social aspects of the person. As a result, the tool does not provide designers with the detailed information they need for each project. In the case of the policy-making strategies, the information gathered in the Quantitative Personas loses its validity with time. In this paper, we propose a new model for the creation of Quantitative Personas. Firstly, we make a review on available surveys regarding disability and aging by comparing them according to their objective, periodicity, the size of the sample, the accuracy and the range of variables that they cover. Secondly, we select variables from the surveys that better describe each aspect included in our model of Elderpersonas. Finally, we propose a model of layered Quantitative Elderpersonas that will be fed from those surveys.

Inclusive Design, Personas, Health Care Design, Policy-making

1. Introduction

Population aging is a serious issue that poses problems for the sustainability of current social policies and healthcare systems. New products and services need to be designed to cope with the demands of an increasingly frail section of the population. The World Health Organization in its Report about Aging and Health (WHO, 2016) highlighted the urgent need to evolve from a model centred on curing acute problems to holistic and integrated care of chronic patients. It seeks to reorient health systems so they better meet the needs of older people.

In order to address these challenges, design tools such as personas are increasingly widely used by policy-making professionals to create more people-centric policies and services. For example, the UK government have created an Open Policy Making toolkit that includes a brief description of personas (Policy Lab, 2013; Cabinet Office, 2016). Another example is the Australian Center for Social Innovation (TACSI) which uses profiles of people (similar to personas) to describe the Baby Boomers that are now reaching retirement age (Burkett and Jones, 2016).

The private healthcare sector has also started using design tools to improve the patient experience. Patient experience can be defined as “the sum of all interactions, shaped by an organization’s culture, that influence patient perceptions, across the continuum of care” (Wolf *et al.*, 2014). The power of personas for healthcare is explicitly described by other authors from the field of healthcare marketing (Leroy, 2016; Piperno, 2016). They consider that it allows the brand and design team to focus their efforts and resources on solutions that will target the right customers, keeping their needs central as the organisation moves closer to products and services. This ensures a seamless experience for customers, who will enjoy improved simplicity, functionality, and usability as a result.

Personas have great potential to be valuable in policy-making and healthcare design. They can help policymakers and designers to understand the complexities of an aging population, and increase their empathy with users. The value of evidence-based personas in this context has been described by various authors. For example, Jones (2013) explains that personas can holistically capture the lived experiences of people, across a range of issues. This is particularly important in healthcare because a wide variety of issues, including life situations and emotional issues, affect people’s ability to engage with care services. Thus personas are an appropriate contribution to “a care-centred design orientation, that can span the different needs of patient, professional and service, and help us define priorities for intervention and redesign” (Jones, 2013). Furthermore, healthcare involves a wide variety of users which can complicate the design process. In this context, personas can help to guide design and policy-making, and inform architects as to the needs and goals of users (Maness, Miaskiewicz and Sumner, 2008). They serve as lively summarizations of user groups, inform design, and inspire problem analysis activities with stakeholders (van Velsen *et al.*, 2012)

However, personas have been criticized for their lack of reliability in decision-making (Tu *et al.*, 2010). To be reliable, they need to be based on accurate and up-to-date data about the population. However, to be useful, they also need to convey the multiple types of information that affect aging and impact policy and healthcare. These include not only medical information but also social, psychological and functional data. As design thinking and HCD practice find their way into processes that aim to cater for patient experience in healthcare there is a growing need to customize methods and techniques originally devised for more traditional applications and domains (Bate and Robert, 2007)

In this paper, we consider how to create personas that are both accurate and holistic. In Section 2 we describe different models of quantitative personas that have been used in the past to describe older or disabled people. Section 3 reviews the available surveys on disability and aging, comparing the variables used in each. Finally, in Section 4 we propose and discuss three different approaches to build on these surveys to create quantitative personas that are more reliable and holistic.

2. Review of personas describing aging

Personas were proposed by Cooper (1999) as a way of helping designers to empathise with users and understand their goals when using a product. They were fictional descriptions of people, each representing a key group of users and their needs and goals. Cooper's original proposal involved using qualitative methods to create the personas. However, in recent publications, several authors have proposed basing personas on quantitative data using statistical techniques (Pruitt and Grudin, 2003; Adler, 2005; Watanabe *et al.*, 2017). These quantitative approaches are always combined with qualitative information to maintain the sense of realism and to create engagement and empathy.

Five sets of personas that consider aging are presented below. Two of them use a qualitative approach (Burkett and Jones, 2016; Gonzalez de Heredia *et al.*, 2017) and the other three also use quantitative data (Marshall *et al.*, 2002; Reeder *et al.*, 2011; Wöckl *et al.*, 2012).

Elderpersonas (Gonzalez de Heredia *et al.*, 2017)

Elderpersonas are a set of personas that aim to convey the diversity among older people through an analysis of the multidimensional aging process. They cover six different kinds of ages: chronological,

biological, functional, psychological, subjective and social age (Cavanaugh and Blanchard-Fields, 2014). As a proof of concept, 36 people were interviewed and the data used create three personas corresponding to successful, normal and pathological aging (Fernández-Ballesteros, 1998). However, these Elderpersonas lack both quantitative data and detail when describing functional age.

Australian Baby Boomers (Burkett and Jones, 2016)

The Australian Centre for Social Innovation published a report of their project Innovating Age in which they described aging as a systemic event, in which the nature of aging is determined by groups of people and by society as a whole, not by individual characteristics. The information contained in each of their four descriptions of people are: family, housing, income, social network, vulnerability factors, major life events and resilience factors. This model also lacks quantitative data and detail about disabilities.

Basic Senior Personas (Wöckl et al., 2012)

The project CURE-elderly-Personas presents 30 basic senior personas that describe the diversity among elderly people from eight European countries. The personas were constructed using data from wave 1 of the Survey of Health, Ageing and Retirement in Europe (SHARE) from 2004. The personas include information related to family status, household size, social activity and contacts, economic situation, limitations, diseases, symptoms and general health. Wöckl et al propose that teams using these personas add further information on goals, frustrations and behaviours specific to each project. This set of personas does not describe how many people each persona represents.

HADRIAN (Marshall et al., 2010)

This computer aided design tool includes a database of 100 real individuals including information on their anthropometry, mobility/capability, disability and coping strategies, as well as a wealth of background data. The individuals were chosen to represent diversity in the adult population. Care was taken to ensure that they cover the range of ages, body sizes and capabilities for certain types of capabilities, but they do not necessarily cover the whole range of social situations or medical conditions.

Modeling the Oldest Old (Reeder et al., 2011)

This project presents two personas based on a cluster analysis of 21 "oldest-old" people, i.e. those aged between 85 and 94. Data was gathered from these people on functional, social, spiritual, medical and cognitive variables. The personas include information on age, education, health conditions, experience with computers and social support. They also contain ratings of general health, functional status and cognition. However, the rating scales used are not very specific. For example, the cognitive status of the personas is described as "minor cognitive changes" and "moderate cognitive changes". The focus on the oldest-old and the small number of personas also limits the applicability of this work.

In summary, personas often include both quantitative and qualitative data. The quantitative data helps to identify the characteristics of different groups and the qualitative data provides an understanding of why those differences exist. Various examples of quantitative personas have been created to describe older people. However, no one example is both holistic and in-depth enough for use in many projects. Policy makers and healthcare managers need up-to-date data and in-depth understanding of people's situations and abilities in order to make new policies, products and services inclusive for everyone. The aim of this paper is to propose approaches for the creation of quantitative personas that give a holistic, deep and up-to-date understanding of the diversity existing among older people and their circumstances.

3. Review of the available statistical data

The currently available statistical data on older people is analysed in this section, with a view to determining if a single database exists with all the data needed to create reliable and engaging quantitative personas. Section 3.1 gives a general description of various different surveys. Section 3.2 compares the range of variables covered by each of these surveys. Finally, Section 3.3 compares the depth with which user capabilities are examined in each survey.

3.1. General description of the surveys

People with disabilities are a minority that has often been neglected when carrying out demographic studies. However, the aging of the population has increased the interest of authorities and governments in understanding the characteristics of this growing group of diverse people. As a result, there has been an increase in the number of national and international surveys related to disabilities and aging. Table 1 summarises the most important surveys identified for this study.

Table 1. Summary of available surveys about disability and aging

N°	Survey	First year	Location	Frequency
1	Washington Group Extended Question Set on Functioning (Washington Group on Disability Statistics, 2011)	2011	60 countries	Varies from country to country
2	World Health Survey (World Health Organization, 2002)	2002	70 countries	Varies from country to country
3	European Health Interview Survey (European Commission, 2006)	2006	31 European countries	Every 5 years
4	English Longitudinal Study of Aging (ELSA) (Marmot <i>et al.</i> , 2002)	2002	England	Every year but longitudinal
5	The Disability Follow-up Survey (Grundy <i>et al.</i> , 1999)	1996	UK	One-off
6	Towards Better Design survey (Tenneti <i>et al.</i> , 2012)	2011	England and Wales	One-off
7	Survey on Disability, Personal Autonomy and Dependency Situations 2008 (EDAD2008) (INE, 2008)	2008	Spain	Every 10 years
8	Elderly people survey IMSERSO2010 (CSIC, 2010)	2010	Spain	2004, 2006, 2010

Table 2. Methodology used in the surveys

	Survey	Objective	Sample	Method	Description of capabilities
1	Washington Group Questions (Washington Group on Disability Statistics, 2011)	Conduct national and local surveys, develop disability specific research	Varies between countries	Interview	Difficulty level/ frequency of problems, intensity, duration
2	World Health Survey (World Health Organization, 2002)	Monitor critical health outcomes and health systems	300,000	Interview	Difficulty level
3	European Health Interview Survey (European Commission, 2010)	Monitor critical health outcomes and health systems	Varies between countries	Interview	Difficulty level
4	ELSA (Marmot <i>et al.</i> , 2002)	Examine the complex relationships and causal processes of aging	10,000	Interview or nurse assessment	Varies from capability to capability
5	The Disability Follow-up Survey (Grundy <i>et al.</i> , 1999)	Help plan welfare support for disabled people	7,200	Expert assessment	Continuum
6	Towards Better Design survey (Tenneti <i>et al.</i> , 2012)	Inform inclusive design	362	Interview + tests	Test results
7	EDAD2008 (INE, 2008)	Inform the National System of Dependency	22,795	Interview	Difficulty level

8	Elderly people survey IMSERSO2010 (CSIC, 2010)	Inform the National Institute of Elderly People	2,535	Interview	Yes or no questions
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In Table 1, we focus on the UK because we consider the tools available in the UK for evaluating inclusion to be a valuable reference. We also include data from Spain to support the development of new tools locally. This work could then be scaled up to apply to other geographical locations

It is difficult to compare disability prevalence and characteristics between countries due to the fact that disability measures and sampling methods vary from country to country (World Health Organization and World Bank, 2011). The World Health Organization have taken steps to address this with the World Health Survey, using the same questions in multiple (though not all) countries. In addition, the Washington Group has proposed tools to help countries carry out surveys in a coordinated way. These include a short and extended questionnaire on the prevalence and levels of disability in the adult population (Washington Group on Disability Statistics, 2011). In addition, they provide a tool to assess disability among children (UN/UNICEF, 2014). These questionnaires are based on the International Classification of Functioning Disability and Health proposed by the World Health Organization (World Health Organization (WHO), 2003)

However, the fact that unified questionnaires have been provided does not mean that every country follows the same process to develop disability statistics. The World Health Survey (survey 2 in the table) and the European Health Interview Survey (survey 3) follow the ethos of the Washington Group because they integrate most of the areas proposed by this group but still they do not ask identical questions. The Disability Follow-up Survey (survey 5) formulates its questions in a similar way, though the details of the questions differ. However, the remaining surveys are different probably due to differences in their objectives as shown in Table 2.

In terms of frequency, not all surveys are conducted on a regular basis. Some are one-offs, others are conducted occasionally, while others are conducted every five years or every year. For personas to be of maximum use in policy making, they need to be kept up-to-date with the latest data. So, ideally, they should be based on surveys that are conducted regularly.

3.2. Range of variables covered by each survey

As described above, there is no single survey that covers all the dimensions involved in aging. Multiple aspects of aging must be considered in order to have a holistic vision of the aging process. The tool Elderpersonas (Gonzalez de Heredia et al., 2017) proposed describing aging using the six kinds of ages presented by Cavanaugh et al. (2002): chronological, biological, functional, psychological, subjective and social age. Table 3 shows which of these each survey covers.

In general, there are two kinds of surveys: one focuses on the prevalence and characteristics of disabilities (surveys 1, 2, 3, 5, 6 and 7) and the other focuses on the wellbeing and circumstances of elderly people (surveys 4 and 8). Many of the first group (surveys 1, 2, 3, 5) describe similar variables using similar scales. In addition, the EDAD 2008 survey from Spain (survey 7) seems to use similar criteria and shares multiple questions with the other surveys, but also includes variables related to daily life, interaction, relationships and health issues. In contrast, the English Longitudinal Study of Aging (survey 4) and the Elderly People Survey (survey 8) focus more on social wellbeing, economic circumstances, family support, participation in society and level of education. Finally, the only survey designed specifically to inform inclusive design is the Towards Better Design survey (survey 6). It uses more specific measures to describe the capabilities needed to interact with everyday products. These different approaches can be identified in the table by the kind of variables that each survey covers or partially covers. If all these types of variables are to be described in a quantitative way, data from different datasets should be combined.

Table 3. Types of variables included in each survey. Y = inclusion, P = partial inclusion.

	Survey name	Range of variables	Chronological Age	Biological Age	Functional Age	Psychological Age	Subjective Age	Social Age
1	Washington Group Questions (Washington Group on Disability Statistics, 2011)	Disabilities, self-care, affect, pain and fatigue	Y		Y			
2	World Health Survey (World Health Organization, 2002)	Disabilities, self-care, affect, pain, fatigue and health	Y	Y	Y			
3	European Health Interview Survey (European Commission, 2010)	Disabilities, self-care, affect, pain and health	Y	Y	Y			
4	ELSA (2002)	Functioning, health, social wellbeing and economic circumstances	Y	Y	P	Y		Y
5	The Disability Follow-up Survey (Grundy et al., 1999)	Disabilities, use of aids	Y		Y			?
6	Towards Better Design survey (Tenneti et al., 2011)	Disabilities, psychological characteristics, product use, anthropometrics and demographics	Y		Y	P		
7	EDAD2008 (INE, 2008)	Disabilities, self-care, daily life, interactions and relationships, health issues	Y	Y	Y			Y
8	Elderly people survey (IMSERSO, 2010)	Family and relationships, loneliness, housing, health and dependency, use of time, wealth, social participation, aging experience, education	Y	P	P	Y	Y	Y

3.3. Comparison of the scales for describing capabilities

Functional ability is particularly important for independent living. For example, the WHO proposes many actions to promote healthy ageing, but all have the same goal: to foster functional ability. This is done in two main ways: by building and maintaining intrinsic capacity, and by enabling those with a decrease in functional capacity to still carry out key activities (WHO, 2015).

In this section we examine how functional ability is considered in the different surveys. This is usually done by examining different kinds of user capabilities, such as vision and mobility. This differs from one survey to another. Surveys differ on the number of questions asked and the scales used to describe each capability. For example, most of the surveys ask between two and four questions about vision (surveys 1, 2, 3, 5) while EDAD2008 (survey 7) asks 20 questions. The additional questions relate to the cause of the disability and the use of assistive technologies rather than the capability itself.

There are also differences in the scales used to measure capabilities. Most of the surveys rate capability according to the level of difficulty experienced when carrying out particular tasks (surveys 1,2,3,7). Possible responses are commonly: No difficulty, Some difficulty, A lot of difficulty, and Cannot do

task. However, in the Towards Better Design survey (survey 6), standardized tests were also performed. For example, vision test charts were used to evaluate different kinds of vision capability.

Interestingly, the questions from the Disability Follow-up Survey were aggregated from levels of difficulty for individual tasks, to a continuum scale for each main capability such as vision or locomotion (Waller, Langdon and Clarkson, 2010). The tasks were organized according to the impact that not being able to do them has on quality of life.

In summary, we conclude that there is no single survey that provides the holistic and detailed view required to inform policy-makers and healthcare designers. If personas are to be created that are effective for policy-making and healthcare, it may be necessary to use data from more than one dataset. However, it is unclear how this can be done in a logic and meaningful way.

4. Methods for creating Quantitative Elderpersonas

We propose three methods that may be effective in creating holistic and detailed personas. Each may be suitable in different situations or have different concerns that need to be worked out. In considering personas, note that each set of personas and the particular choice of information included in it should be informed by the purpose for which the personas will be used.

4.1. New Survey(s)

The most complete way to create quantitative elderpersonas is to create a new detailed survey that includes all the data needed to describe the elderly population in a holistic and detailed way. This survey would include questions from various other surveys, and would cover the following six areas:

1. Demographic data, e.g. age, sex, town size, whether they live alone or with someone else.
2. Biological age, e.g. health conditions, general health and clinical history.
3. Functional age, e.g. level of functional capabilities (related to daily activities and product use), pain and fatigue and how these affect daily activities.
4. Psychological age, e.g. energy, fatigue, levels of motivation.
5. Subjective age, e.g. whether they consider themselves old or not.
6. Social age, e.g. social network, level of support, participation in social activities.

Ideally, the survey would be conducted with a sample that was representative of the whole population in the country of interest. It is also important to determine the frequency at which the survey would be repeated in order to provide updated data to inform policy makers and healthcare managers. Dependency evaluation surveys are usually repeated every year. Coordinating with the organisations that develop these evaluations would provide one way to manage the survey more easily. After gathering the data, a technique such as cluster analysis could be used to group it into clusters that represent groups of people in the population. A persona could then be created to represent each group.

This method would create the most reliable set of quantitative personas, but would be very expensive to put into practice. Another approach is to include a short set of identical questions in more than one survey. These questions could cover the six key areas listed above briefly. It may be easier to add a few questions to a survey that would be conducted anyway than to conduct a whole new survey.

The resultant datasets would then be linked by having some common variables. This would provide the potential for linking data from one dataset that provides (for example) functional data with data from another dataset providing social data. This would give a more solid base for personas than using two separate unlinked surveys. However, the analysis required to do this linkage may not be straight forward.

4.2. Layered Personas

The second proposal is to combine information from more than one survey using layered personas (Marcengo et al. 2009). These personas consist of several layers, different combinations of which can

be used in different projects. Some of the layers are developed once and reused in multiple projects, and some are developed for specific projects, taking into account the application context of individual projects. Different layers can also provide different kinds of information relevant to different situations.

A similar structure could be used to present the quantitative information gathered from different sources. For example, one layer could include information on functional age from one survey, while another includes information on social age from another.

However, there are challenges over how the layers from different surveys could be combined into a coherent set of personas. One potential way to do this is to first create clusters from the different surveys separately. An example is shown in Figure 1. Clusters A1 to A3 are produced by cluster analysis on survey A, and contain information on functional age. Clusters B1 to B4 come from survey B and contain information on social age. Both sets of clusters also contain some information on demographic variables.

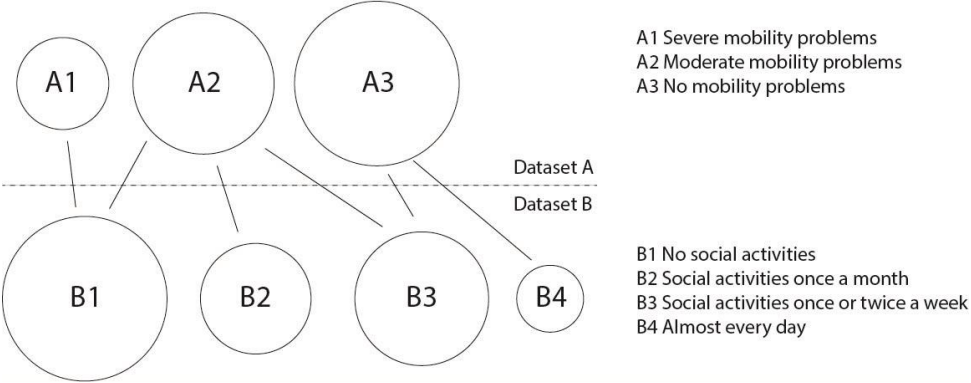


Figure 1. An example of linking between clusters in different datasets

To combine the sets of clusters, information is needed about how functional and social age are related. This could be obtained from interviews or by using a separate survey that examines both functional and social age, even if in less depth. Using linked surveys (as proposed in Section 4.1) may also help.

As an example, Table 4 shows some hypothetical results for how one aspect of functional age could be related to one aspect of social age. This could be drawn from a different survey or source than Survey A or B. It indicates that people with severe difficulty in moving about the house tend to participate less in social activities, and people with no difficulty tend to participate more. People with moderate difficulty are fairly evenly split between participation and no participation.

These relationships can be drawn on the cluster diagram as shown in Figure 1. The A1 cluster represents people with severe movement issues, corresponding to severe difficulty moving about the house. Table 5 indicates that in general these people tend not to participate in social activities (cluster B1). So a line is drawn between the two. In some cases, one cluster in the top row is joined to more than one cluster in the bottom row. Each pair of clusters is then expanded into a persona. For the example in Figure 1, personas would be created for: A1-B1, A2-B1, A2-B2, A2-B3, A3-B3 and A3-B4.

Table 4. Example of cross-tabulation of one functional variable (difficulty moving round the house) with one social variable (participation in social activities).

	Participation in social activities	No participation in social activities
Severe difficulty moving around the house	20%	80%
Moderate difficulty moving around the house	60%	40%
No difficulty moving around the house)	90%	10%

This proposed approach is still tentative and has many challenges. When the information used for pairing the clusters is more crude than the information in the clusters themselves, it can be hard to decide on the exact pairing. This is shown in the example above, where Table 5 only provides information on whether or not there is participation in social activities, not about the level of that participation. The person creating the clusters has to use some judgement when creating the pairings. In addition, it is unclear how well the pairing would work when the clusters contain information on more than one variable. There is also a danger that this method would result in a very large number of clusters that are hard to manage.

Another issue is that the pairing process makes it difficult to determine the size of the resultant clusters. One advantage of using survey data is that it provides an estimate of how large each cluster is (e.g. A1 to A3 and B1 to B4 in the example above). However, it is unclear how large the paired clusters are (e.g. the pair A2-B2). Further research could investigate ways to address this problem.

4.3. Iterative Personas

Another way to use data from more than one survey is to base the personas on a small number of real people. Data can be gathered about these real people using measures from both surveys. An iterative process can be used to refine and expand this set, and ensure that it is representative of the whole population. The process is summarised in Figure 2 and described in more detail below. This description assumes two surveys (Survey 1 and Survey 2) but it may be possible to expand the process to a larger number of surveys.

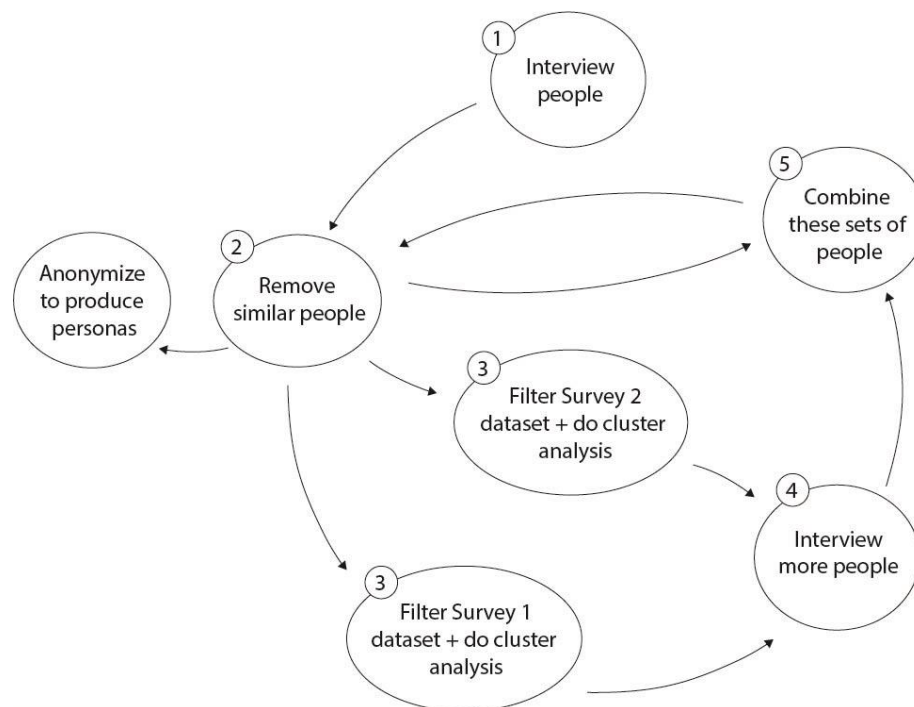


Figure 2. The development process for iterative personas

1. Recruit some people who are relevant to the decision that needs to be made. A reasonable number of people (n) might be 10 to 20. Give them the relevant survey questions from Survey 1 and Survey 2. It may also be helpful to interview them or conduct user trials with them to get more in-depth information about them on topics of interest to the project.
2. Examine this set of n people and filter out people who are very similar to each other. For person P1 and person P2 to be considered similar, they should have similar results on all the measures of interest from both Survey 1 and Survey 2. The measures of interest should be chosen based on the purpose that the personas will be used for. If P1 and P2 are similar based

- on the measures in Survey 1 but not those in Survey 2, keep them both in. If they are similar on both measures, keep one of them and remove the other. This gives a set of N people.
3. Use these N people as a starting point for the persona set. Filter the Survey 1 dataset to exclude all participants who are close to one of these N people (based on their answers to the Survey 1 questions). Do the same for the Survey 2 dataset based on the answers to the Survey 2 questions. For each of the N people, calculate how many participants from Survey 1 are similar to that person, and how many participants from Survey 2 are similar to that person. This gives a rough indication of how many people that person represents. Perform cluster analyses on the filtered datasets from Survey 1 and Survey 2. These will indicate the kinds of people that are currently not included in the persona set.
 4. Recruit M additional people that roughly match the clusters from Survey 1 and that roughly match the clusters from Survey 2. Give all of them the (relevant) survey questions from both Survey 1 and Survey 2, and any additional interviews or user trials desired.
 5. Ideally, these M people would match the centres of all the remaining clusters, but this is unrealistic in practice. Often it is hard or impossible to find exactly the right people. As a result, it is useful to iterate the process. Add the M additional people to the previous set of N , then return to step 2 to check how well the new set of people covers the whole population, and recruit additional people if necessary.

In theory, as this process is repeated multiple times, the set of people who have been interviewed will cover an increasing proportion of the total population. At some stage it will be necessary to make a judgement call as to when the set is comprehensive enough to inform the decision that needs to be made.

Note that this process produces a set of real people with rich data. An anonymization step may be required to translate these into personas that can be shared and used more publicly. For example, Jay, Harper and Calman (2012) created personas from descriptions of real people, but changed contextual data that might identify them, such as specialist hobbies and occupations, and details of their children.

5. Discussion

Creating a new survey that covers all areas of interest (Section 4.1) would be the best approach in terms of the quality of the data. However, this would require large amounts of resources and time. In addition, in order to use public resources efficiently, coordination with different institutions that carry out similar surveys would be necessary. This coordination may be difficult due to institutions' different objectives.

Section 4.1 also proposed an alternative in which a small set of questions is added to a range of existing surveys. This would be more feasible, and would be a useful way of obtaining up-to-date data. However, it may still involve difficulties in terms of persuading and collaborating with different organisations. This method does not solve the issue entirely. Although the surveys would provide data on the whole range of variables, there would still be no survey that covers the whole range in detail. However, it would provide a useful way to link together data from different surveys, perhaps using the method suggested in Section 4.2.

Section 4.2 proposed using a layered approach. It suggested creating clusters from more than one survey and then matching them based on data from interviews or from surveys (such as those from Section 4.1). However, matching the clusters from the different datasets could present difficulties, particularly if the clusters are based on many different variables. It may also be difficult to maintain the quantitative nature of the personas and determine how many people are represented by each combination of clusters. This may depend on the data available for the matching. This proposal is still in early stages. Further work is needed to try it out on real datasets and examples. Detailed statistical analysis is required to determine how effective the cross-tabulation could be and whether it is possible to determine the size of cluster combinations.

The method proposed in Section 4.3 used profiles of real people which are matched to multiple surveys based on their answers to those survey questions. This proposal gives a good mix of quantitative and qualitative data, and the iterative process provides a way to gradually increase the quality of the personas. It may be more difficult to implement than the proposal in Section 4.2 due to the need to interview additional people, but it is substantially easier than creating a whole new survey.

All of these methods have some potential. The third one seems particularly promising. However, all of these are early-stage proposals. Further work is needed to try them out in practice and see whether and how they work out.

To consider the feasibility of developing a proof-of-concept for iterative personas, a hypothetical scenario was considered where personas would be developed to inform the provision of transport services to day care facilities. So this scenario, various types of data would be relevant, including the levels of mobility and independence (functional age), levels of support (social age) and motivation (psychological age). Referring to Table 3, no single survey covers all of these types of data, but EDAD2008 (survey 7) covers functional age and social age, while IMSERSO 2010 (survey 8) covers psychological age and social age.

Furthermore, data is available from 36 previous interviews (Gonzalez de Heredia et al., 2017) that cover all of these types of data. These interviews could be used as the starting point for the iterative process described in Section 4.3 and Figure 2. Firstly, interviewees would be filtered to remove those that are essentially duplicates, in terms of the relevant factors for providing transport services to day care facilities (step 2). Therefore, for any interviewees with similar levels of mobility and similar levels of independence, and similar levels of support and similar levels of motivation, the analyst would choose only one interviewee to keep. The data from EDAD2008 and IMSERSO2010 could then be filtered to remove survey participants who were similar to these interviewees, and perform cluster analysis on the remaining survey participants (step 3). These clusters would describe the kinds of people that are 'missing' from the current known set of interviewees, which can inform the recruitment of further interviewees (step 4). This additional data would then be combined with the previous set of interviewees (step 5), before repeating the process again from step 2. This iteration would continue until the analyst judges that the set of known interviewees is comprehensive enough to inform the intended purpose, in which case the known interviewees would be anonymised to produce the persona set.

6. Conclusions

We have identified an issue with using current personas in policymaking and healthcare design, due to the lack of a single database that represents the aging population in a holistic and detailed way. We have proposed three different methods to deal with this that enable the production of more effective personas. The first method involves collecting new survey data, the second matches clusters from different surveys, and the third uses a small sample of real people to link data from different surveys. All three methods have potential but all require further work to refine and test the proposals. We present this paper to the conference with the aim of generating discussion on this topic and on the proposals.

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