Rural and non-rural digital divide persists in older adults: Internet access, usage, and attitudes toward technology

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Abstract

Background: The digital age divide remains persistent despite the recent increase in internet use among older adults. Additionally, older rural residents are at greater risk of being digitally disconnected.

Objective: Guided by the social determinants of health framework, our study aims to examine how one’s residential area relates to (1) internet use, (2) subtypes of usage patterns, and (3) attitudes toward technology use in later life.

Method: Cross-sectional data were drawn from the 2012 Health and Retirement Study. The analytic sample consisted of 1,566 older adults aged 50 and above. Chi-square tests and logistic regression analyses were performed.

Results: Internet access rates were significantly lower in rural residents (54%) compared to the urban (66%) or the suburban group (61%). Compared to urban residents, those residing in suburban areas were less likely to use health technology while those living in rural areas had lower odds of communication, financial, and media technology use. Furthermore, the association between urban-rural residence and attitudes toward technology was compared among non-users (N = 633). Older adults in rural areas showed more unfavorable perceptions of technology than urban residents. They were less likely to conceive technology as “easily available,” but more likely to perceive it as “too complicated” and “too hard to learn.” No significant differences were found between rural and suburban residents.

Conclusion: Our findings suggest that older adults in rural areas, notably lag in using and adopting digital technology. Comprehensive intervention efforts are needed to narrow the digital divide for rural communities.

Keywords: technology, digital divide, rural, urban, attitude, older adults

INTRODUCTION

With the pervasive spread of technology and digital technologies used in every aspect of life, research has noted the development of a digital divide (Ball et al., 2019; van Dijk, 2006). As elaborated through these studies, a digital divide refers to a division in access to digital technologies for certain populations. More specifically, van Dijk and Hacker (2003) proposed four access-related gaps that might lead to a digital divide. The first was a mental gap, which refers to emotional or psychological gaps that emerge from people’s lack of experience with digital technologies. Next was the material access gap, referring to the more traditional lack of access, such as lack of access to technologies. Third, they referred to a lack of skills, referring to how experience with technologies would increase an individual’s technology skills. Lastly, they mention the usage gap, referring to differences in usage that lead to various usage patterns (such as simpler communication tasks only vs. more complex, social capital building tasks, for instance). Despite these multiple aspects of digital technology use, most previous studies on the digital divide have been primarily focused on accessibility (whether the participants can use the technology or not), leaving usage patterns and personal attitudes underexplored.

The digital divide literature has conceptualized three levels of the digital divide: (1) the first-level divide focusing on disparities in Internet access; (2) the second-level divide referring to unequal Internet skills and usage; (3) the third-level divide related to the outcomes of Internet use (van Dijk, 2020). While more recent research attention shifts towards the second-and third-level divide as mobile broadband becomes nearly ubiquitous, the first-level digital divide still calls for further investigation in the context of socio-
economic disparities (Gonzales, 2016). On the other hand, theoretical frameworks have suggested two distinct reasons for the non-use of the Internet. First, a lack of material resources may push individuals involuntarily away from Internet use (Livingstone & Helsper, 2007). Second, psychological factors such as motivation, attitudes, and interest may drive individuals’ voluntary decisions not to use digital technologies (van Dijk, 2005). Studies found that inequalities in Internet access are likely to be attributable to involuntary exclusion, while disparities in Internet skills and usage are also shaped by personal motivations and attitudes (Eynon & Helsper, 2011; Yu et al., 2016). The current study aims to investigate the role of one’s residential area in relation to the first- and second-level digital divides using a representative sample of older Americans. We consider internet access and specific types of technology usage as outcome variables. In addition, we focus on non-users of communication technologies to examine what contributes to their attitudes toward technology.

It is important to note that the digital divide is nuanced and shows different patterns of change depending on which aspects are considered: some digital gaps are narrowing down over the last few decades while others remain. For example, the rate of Americans who don’t use the internet has decreased rapidly from 48% in 2000 to 7% in 2021 (Perrin & Atske, 2021). Similarly, recent studies found that the vast majority of people have a smartphone or technological divide (Barrantes & Vargas, 2017). While the general access to the internet or technology has been improved, individuals with low-income, rural residents, and minorities have a limited number of digital devices for online access and rely primarily on smartphones (Vogels, 2021a). They are also less likely to have broadband internet at home, creating other disparities such as educational and health inequalities (Ong, 2020; Vogels, 2021b).

The digital divide leads to the exclusion of populations based on different factors. Age has been a major leading factor in the digital divide, placing older adults into one of the largest affected populations (Tsai et al., 2015). As of 2019, 27% of Americans aged 65 and older and 12% of adults aged 50 to 64 were not using the Internet, which is a sharp contrast to the nearly ubiquitous internet usage among young adults; 100% of 18–29-year-olds and 97% of 30–49-year-olds reported using the internet (Pew Research Center, 2019). Although one of the largest populations affected by the digital divide is older adults (Tsai et al., 2015), digital technologies may be most beneficial to this population. Older adults’ technology use spans various avenues of life. Using information and communication technologies (ICTs) has shown to have increased feelings of mattering among older adults (Francis et al., 2019), connecting them with loved ones (Heo et al., 2015; Quan-Haase et al., 2017; Sum et al., 2008), and assisting with managing their health through medical and health-related technologies (Levine et al., 2016; van Deursen, 2020). With increasing medical treatment improvements, older adults account for large percentages of both the U.S. and global populations. By 2030, 20% of the U.S. population will be older adults – the largest population of older adults in U.S. history (Centers for Disease Control and Prevention, 2013). Given the beneficial effects of technology use and demographic changes, more research efforts are needed to close the older population’s digital divide.

In addition to the generally low technology adoption rates among older adults, it is vital to understand the within-group heterogeneity. Those with lower socioeconomic status and poor health have been reported to be at greater risk of being excluded from technology use (Gell et al., 2015; Silver, 2014). Amongst various predictors of older adults’ technology use, a recent growing body of literature has shown that rural living is significantly associated with limited access to the internet and electronic devices (Berner et al., 2015; Calvert et al., 2009). Moreover, in qualitative studies, older adults living in suburban or rural areas reported negative perceptions of using technologies, such that lack of knowledge about how technologies function or difficulty in learning how to use them greatly limited their use (Marston et al., 2019; O’Brien et al., 2014). Prior studies suggested low rural residents’ socioeconomic status levels partly contributing to the urban-rural differences in technology use (Hale et al., 2010). However, underdeveloped infrastructure and broadband services in rural communities, which are related to high-speed internet access, can also result in their residents’ limited technology use (Anderson, 2018; Greenberg et al., 2018; Korupp, 2005). Indeed, having access to technology, particularly high-speed internet, is a critical facilitator of older adults’ technology use (Hanson, 2010; Marston et al., 2019).

The observed digital divide by age and residential area is well encapsulated in the social determinants of health framework (World Health Organization [WHO], 2010), which posits that the source of health inequities originates from “the conditions in which people are born, grow, work, live, and age, and the wider set of forces and systems shaping the conditions of daily life.” (WHO, 2020, p.1). According to the framework, the digital divide could occur due to broader systemic and socioeconomic contexts and individual-level sociodemographic factors (i.e., age and gender). Thus, we argue that a geographic
residence where older adults live would be an important social determinant of digital technology use by offering different opportunities and challenges embedded in regional infrastructure and community support.

The present study
Although prior studies suggested some valuable insights into the urban-rural digital divide for the older population, most relied on small convenient samples or measured a single aspect of technology use (i.e., internet access). In addition, there were few studies focused on different usage patterns or psychological factors. The current study aims to investigate whether one’s residential area relates to (1) their internet use, (2) the subtypes of their technology usage patterns, and (3) their attitudes toward technology use. We conducted a series of regression analyses on data of a nationally representative sample of U.S. adults aged 50 or older. Building on the previous literature and the social determinants of health framework, we hypothesized the following:

Hypothesis 1: Older adults living in suburban and rural areas will use the internet less than their urban counterparts.

Hypothesis 2: Older adults living in suburban and rural areas will be less likely to use different technologies (communication, financial, health, and media) than urban residents.

Hypothesis 3: Older adults living in suburban and rural areas will have more unfavorable attitudes toward technologies than those residing in urban communities.

Methods
Data and sample
The Health and Retirement Study (HRS) is a nationally representative panel survey of approximately 20,000 Americans aged 50 or older. The HRS has been repeated biannually since 1992. Participants are asked a wide range of questions on aging (e.g., health, jobs, retirement, and social relationships) through face-to-face or telephone interviews. The HRS’s sampling design, survey procedures, and questionnaires are described in detail at the HRS website (http://hrsonline.isr.umich.edu). We retrieved the respondent’s socio-economic and health information from the RAND HRS data file, which is cleaned and processed by the RAND Center for the Study of Aging (Santa Monica, California). The RAND HRS data set provides imputed values for income and health status that were generated using all information available with a consistent imputation method.

In 2012, a randomly selected subsample of participants was administered to an experimental module that included questions about technology use and perceptions of the barriers and benefits. A total of 1,620 community-living adults aged 50 or older completed the module. We excluded 54 respondents (3.33%) who had missing data on covariates, internet access, and internet usage variables, which resulted in an analytic sample of 1,566 individuals. Among the respondents, those who did not use any communication technologies such as email and social network sites (N = 633) were further asked about their attitudes toward technology. We use this subsample of respondents for the analyses of non-users’ attitudes toward technology. Ethical approval for the HRS was obtained from the University of Michigan Institutional Review Board and the National Institute on Aging. All respondents have given their written informed consent before the data collection.

Measures
Internet access
Participants’ internet access was assessed with a single item: “Do you regularly use the Internet (or the World Wide Web) for sending and receiving email or for any other purpose, such as making purchases, searching for information, or making travel reservations?” Responses were coded as 1 = yes or 0 = no.

Technology usage
Technology usage was measured by a checklist of various kinds of electronic technologies. The list included fifteen items (1 = yes for each), which can be grouped into four broad types of technology: (1) communication technology (i.e., email, social networks, online calls, online chatting, and smartphone); (2) financial technology (i.e., online bill payment and online banking); (3) health technology (i.e., online wellness program, online health information seeking, devise use of health monitoring, and Wii fit use); and (4) media technology (i.e., e-readers/tablets, MP3 players, live-streaming radio, T.V., or movies, and video games). Responses were dichotomized for each technology category, where 0 = non-user and 1 = users who checked one or more technology items in a given category.

Attitudes toward technology
Participants who reported no use of any communication technologies, or non-users, were further asked about their attitudes toward technology. They answered yes or no to eight questions: (1) if they would be interested in trying any communication technologies, (2) whether technology is too expensive, (3) easily available, (4) too complicated, (5) too hard to learn, (6) takes too much time to learn, (7) if they are opposed to learning new technologies, and (8) whether it is difficult to keep up with changes in technology.

Urban-rural residence
The residency was categorized as urban, suburban, and rural using the 2013 Beale Rural-Urban
Continuum (RUC) codes. Developed by the S.U.S. Department of Agriculture’s Economic Research Service, the RUC codes represent the degree of urbanization by population size and adjacency to metropolitan areas. Urban refers to metropolitan counties with a population size of 1 million or more (RUC code 1). Suburban indicates metropolitan counties of 250,000 to 1 million population (RUC code 2). Finally, rural areas include counties with fewer than 250,000 population, both adjacent or not adjacent to a metropolitan area (RUC code 3–9).

Covariates
Age (in years), gender (0 = male; 1 = female), race/ethnicity (0 = non-Hispanic White; 1 = racial/ethnic minority [non-Hispanic black, Hispanic, American Indian, Alaskan Native, Asian, and Pacific Islander]), and marital status (0 = separated, divorced, widowed, or never married; 1 = married or partnered) were included as demographic variables. Educational attainment (in years) and annual household income served as indicators of socioeconomic status. Education was assessed by the number of years of formal schooling completed (range: 0-17), and annual household income included income from all sources such as earnings, pensions, and social security. Because the distribution of annual household income was highly skewed, log-transformed values were used for multivariate analyses. As for indicators of health conditions, chronic diseases, functional limitations, self-rated health, and depressive symptoms were considered. Participants were asked to report whether or not they had any of the following chronic diseases diagnosed by a physician: high blood pressure, diabetes, cancer, lung disease, coronary heart disease, psychiatric problems, arthritis or rheumatism, and stroke. The number of chronic diseases was calculated by summing up all specific diseases (range 0-8). Functional limitation was assessed by asking participants’ difficulties in performing activities of daily living (ADLs) such as bathing, dressing, walking across a room, and getting in and out of bed. The composite scores were dichotomized into 0 = no functional limitation and 1 = one or more functional limitations. Self-rated health status was measured with a single item, “Would you say your health is excellent, very good, good, fair, or poor?” Responses were coded as 1 = poor and 5 = excellent so that higher scores indicated better subjective health status. Depressive symptoms were measured with the 8-item version of the Center for Epidemiological Studies Depression (CES-D) scale (Andresen et al., 1994; Radloff, 1977). Participants were asked to report whether they have experienced eight depressive symptoms (e.g., lonely, feeling depressed, and sad) in the past week. Response options were 0 = no or 1 = yes. After reversing two positive mood items, total scores were computed by adding responses for each item. The CES-D scale’s internal consistency was high in our sample, with the Kuder-Richardson 20 (KR-20) coefficient of .82.

Analytical strategy
Descriptive statistics were conducted to review the sample’s demographic, socioeconomic, and health characteristics by urban-rural residence. Comparisons were made using adjusted Wald or chi-square statistics. Next, group differences in internet access and usage were compared across urban, suburban, and rural residents through chi-square tests. Separate multivariate logistic regression analyses were performed for each technology used to examine their association with rural-urban residence (N = 1,566). Similarly, we used multivariate logistic regression models to investigate the association between rural-urban residence and attitudes toward technology among non-users (N = 633). For the analysis of non-users’ attitudes toward technology, the pairwise deletion was applied to make optimal utilization of available data (Ns = 556–629). We used sample weights to adjust for differential
Urban/rural digital divide among older adults

Table 1 summarizes the sample characteristics by rural-urban residence. The sample was comprised of 52% urban, 23% sub-urban, and 25% rural residents. Compared to urban residents, the rural group included a lower proportion of racial/ethnic minorities and those with functional limitations. In addition, rural residents had lower annual household income levels and poorer self-rated health than their urban counterparts. Both suburban and rural residents had fewer years of education and reported more chronic diseases compared to the urban group. No significant group differences were found for age, gender, marital status, and depressive symptoms.

Urban-rural group differences in internet access and technology usage

Table 2 presents a cross-tabulation of internet access and different types of technology usage by urban-rural residents. About 66% and 61% of urban and suburban residents were regular internet users, whereas only 54% of rural residents regularly used the internet. In terms of communication technology use, significantly fewer individuals in the rural group (60%) reported use of at least one communication technology than their urban counterparts (71%). For specific technology, significance was found in email use (55% in rural vs. 69% in urban), online chatting (17% in rural vs. 23% in urban), and smartphone use (16% in rural and 24% in sub-urban vs. 32% in urban). Compared to urban residents, suburban and rural residents had fewer individuals who used one or more financial technologies (38% in rural and 43% in sub-urban vs. 52% in urban). About a third of suburban and rural residents used online bill payment, while more than 40% of urban residents did so. Rates of online banking users were also significantly lower among the rural group than their urban counterparts. Similarly, there were fewer individuals from the suburban group who used at least one health technology than the urban group. More urban residents sought online health information than the other two groups (57% in urban vs. 43% in sub-urban and 44% in rural). Finally, the rural group had significantly fewer media technology users (37%) compared to the urban group (53%). Specifically, significant differences between urban and rural residents were found in all sub-categories, whereas the differences between urban and suburban were only found in the use of MP3 players and live-streaming radio, TV, or movies.

Logistic regression models of technology usage

We further examined the observed group differences in technology usage with multivariate logistic regression analyses. Table 3 shows the association between rural-urban residence and four different technology use types after controlling for demographic, socioeconomic, and health covariates. Compared to urban residents, rural residents had lower odds of using all technology types except health technology. On the other hand, the suburban group showed decreased odds of using health technologies than the urban group. No significant associations were found in the use of communication, financial, and media technologies.

Logistic regression models of non-users’ attitudes toward technology

Table 4 presents the models predicting attitudes toward technology among non-users of communication technology. We found that three particular attitudes toward technology were significantly associated with the rural-urban residency. Compared to urban residents, rural residents were less likely to report that technology is easily available and more likely to think that technology is too complicated and too hard to learn. It was noteworthy that no significant differences were found between suburban and urban residents in their attitudes. The associations of the rural-urban residence with other five types of attitudes...
toward technology (i.e., have interest in trying technology, technology is too expensive, takes too much time to learn, too difficult to keep up with changes, and if they are opposed to learning new technologies) were not significant and thus not shown in the table for simplicity.

**Discussion**

Informed by the social determinants of health framework, we analyzed the HRS data to explore whether one’s residence (urban, suburban, or rural) is associated with less use of the internet, limited usage patterns of different technologies, and more negative attitudes toward technology in later life. Our bivariate results indicate that individuals in rural communities use the internet at a lower rate (54%) as compared to urban populations (66%). Furthermore, rural living older adults used technology at lower rates for different technology usage sub-types, including communication, financial, and media technologies. Compared to urban residents, older adults living in suburban areas also reported lower usage rates of certain types of technology use, such as smartphones, online bill payments, and online health information seeking.

Our multivariate models of technology usage further showed that rural residency (vs. urban) was significantly associated with decreased odds of using communication, financial, and media technology, while suburban residency (vs. urban) was related to lower odds of health technology use. Overall, these findings add to the body of past research, pointing out a digital exclusion of rural populations in general (Greenberg et al., 2018; Perrin & Duggan, 2015) and in the older populations (Berner et al., 2015; Calvert et al., 2009). Importantly, our analyses on various usage patterns of ICTs extend the scope of the existing literature where a single aspect of technology use (i.e., internet use) was often focused.

Our findings documented that older rural residents particularly do not engage in online communication and media. Such exclusion could lead to many disadvantages, including connection with loved ones and personal mattering. This mattering and connectedness often occur through the usage of ICTs, which allow for the maintenance of social connectedness and receiving social support from their networks (Francis et al., 2019). Mattering has been explained as occurring due to social interactions among people, which helps reinforce their value to others (Fazio, 2009), but in situations where in-person interactions may not be possible, especially social interactions that reinforce ones’ value to others (Fazio, 2009). Therefore, in lieu of and in addition to face-to-face interactions, using ICTs may be one way to enhance a sense of mattering among older adults and potentially improve their general well-being. This is even more relevant in times like the current COVID-19 pandemic, where people are required to socially distance and avoid social interaction, especially for older adults who are more susceptible to the illness. We also found lower usage of technology for financial reasons among rural residents. Today, most financial institutions allow consumers to perform tasks such as banking and bill payments online, making transactions easier, quicker, and even safer. Therefore, those residing in rural areas are more likely to lose all these advantages of being digitally connected.

Additionally, technology is being used in the health sector in every way today. Following our findings, older adults in both suburban and rural areas use ICTs at a lower level for health purposes (e.g., online health information seeking) than their urban counterparts. This is unfortunate, considering that the benefits of health technologies may be particularly pronounced for non-urban communities. For example, telehealth can provide healthcare services with no traveling and maximizing efficiency (Heinz et al., 2013). Furthermore, Ramisetty and Adams (2020) describe how social determinants of health play a role in the development of the digital divide.
Urban/rural digital divide among older adults

Table 4. Logistic regression models of attitudes toward technology among non-users (N=633)

<table>
<thead>
<tr>
<th>Variables</th>
<th>Easily available (n = 599)</th>
<th>Too complicated (n = 611)</th>
<th>Too hard to learn (n = 599)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Rural-urban residence (Ref: urban)</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Sub-urban</td>
<td>0.91 (0.46–1.80)</td>
<td>1.07 (0.62–1.86)</td>
<td>1.34 (0.77–2.35)</td>
</tr>
<tr>
<td>Rural</td>
<td>0.47 (0.26–0.85) *</td>
<td>1.81 (1.02–3.21) *</td>
<td>1.93 (1.13–3.30) *</td>
</tr>
<tr>
<td><strong>Demographic characteristics</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Age (in years)</td>
<td>0.99 (0.97–1.01)</td>
<td>1.03 (1.01–1.06) **</td>
<td>1.05 (1.02–1.07) ***</td>
</tr>
<tr>
<td>Female</td>
<td>0.64 (0.37–1.11)</td>
<td>1.23 (0.77–1.97)</td>
<td>0.89 (0.56–1.41)</td>
</tr>
<tr>
<td>Racial/ethnic minority</td>
<td>0.48 (0.28–0.84) **</td>
<td>1.10 (0.62–1.96)</td>
<td>0.77 (0.45–1.33)</td>
</tr>
<tr>
<td>Married/partnered</td>
<td>1.83 (1.10–3.07) *</td>
<td>1.13 (0.66–1.93)</td>
<td>1.09 (0.67–1.78)</td>
</tr>
<tr>
<td><strong>Socio-economic status</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Educational attainment (in years)</td>
<td>1.09 (1.01–1.17) *</td>
<td>0.83 (0.75–0.93) ***</td>
<td>0.83 (0.76–0.91) ***</td>
</tr>
<tr>
<td>Annual household income</td>
<td>1.06 (0.91–1.24)</td>
<td>0.81 (0.58–1.14)</td>
<td>0.88 (0.70–1.11)</td>
</tr>
<tr>
<td><strong>Health conditions</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Chronic diseases (numbers)</td>
<td>0.95 (0.78–1.15)</td>
<td>1.13 (0.94–1.35)</td>
<td>1.05 (0.88–1.15)</td>
</tr>
<tr>
<td>Functional limitation</td>
<td>0.75 (0.39–1.45)</td>
<td>2.00 (1.00–4.01)</td>
<td>1.83 (0.95–3.52)</td>
</tr>
<tr>
<td>Self-rated health</td>
<td>1.07 (0.79–1.44)</td>
<td>1.02 (0.77–1.34)</td>
<td>0.96 (0.73–1.24)</td>
</tr>
<tr>
<td>Depressive symptoms</td>
<td>1.07 (0.94–1.21)</td>
<td>1.06 (0.92–1.23)</td>
<td>1.22 (1.07–1.39) **</td>
</tr>
</tbody>
</table>

*p < .05; **p < .01; ***p < .001.
Note: Estimates were weighted. Racial/ethnic minority = non-Hispanic black, Hispanic, American Indian, Alaskan Native, Asian, and Pacific Islander.

(even in the context of COVID-19) and can reinforce inequity based on social factors, including health and healthcare access (along with education and economic stability). The pandemic has further increased how technology is being used for healthcare, and older adults can now get some of the care they need inside their homes – exclusion through the digital divide is likely to increase adverse health outcomes.

Our analyses of non-users’ attitudes toward technology deserve particular attention. We showed that older adults in rural areas were more likely to perceive that technology is less easily available, too complicated, and too hard to learn as compared to the urban group. These findings are in line with previous qualitative work where older suburban or rural residents reported that their use of technologies was limited due to lack of knowledge about technologies or difficulties in learning (Marston et al., 2019; O’Brien et al., 2014). However, there has been a dearth of quantitative studies, particularly with a nationally representative sample, that examined perceptions or attitudes toward technology use among older adults living in non-urban communities.

Overall, the present study findings support that older adults living in suburban and rural areas not only have lower access to the Internet (Hypothesis 1), but they also use fewer Internet-based technologies (Hypothesis 2) compared to those from the urban. Thus, our work demonstrates the presence of the digital divide’s first level (inequal access to the Internet) and second level (different technology skills and usage) in the context of geographic location. Furthermore, we showed that non-users of communication technologies from suburban and rural areas have more unfavorable attitudes toward technologies than those in urban communities, supporting Hypothesis 3. Importantly, our results pertain to the conceptual discussion on the digital divide (Gonzales, 2016; van Dijk, 2005, 2020). According to van Dijk and Hacker (2003), the digital divide is not only related to the simple lack of access but also understood as usage patterns or psychological gaps, as shown in our study. In particular, rural non-users’ more negative attitudes towards technology may serve as critical determinants of their voluntary decision not to use technology. Therefore, more research efforts should be paid to explore multiple aspects of the urban-rural digital divide.

Limitations

Despite our best efforts, we acknowledge certain limitations of our research. First, the HRS dataset is from 2012, so it is possible that this information could be a little outdated as compared to current technology usage patterns for older adults. For similar reasons, the dataset does not include more recent technologies such as wearable trackers, for instance. Critically, our study utilized cross-sectional data, which does not allow for making causal inferences. Future studies should use a longitudinal research design to provide a complete understanding of the relationship between one’s residence and technology use in later life. In addition, although we included a wide range of information on specific types of technology use, data on frequency or duration of use was not obtained. Based on our findings, a next step forward may include those variables to examine whether the urban-rural residence relates to how long or frequently one uses digital technologies.

Conclusions

Despite these limitations, however, we believe that the results of our study provide valuable in-
This work was supported by the Ministry of Education and the National Research Foundation of Korea (NRF-2018S1A3A2075402). Funding was also provided by the National Institute on Aging (NIA U01AG009740) and the Social Security Administration. Data was from the Health and Retirement Study, which was funded by the National Institute on Aging (NIA U01AG009740) and the Social Security Administration. The current study was not preregistered. Data, analytic methods, and study materials will be made available upon request through the corresponding author.

Conflict of interest
The authors declare that there is no conflict of interest.

Author statement
Hee Yun Lee: Funding acquisition, Conceptualization, investigation, methodology, writing – review & editing. Shaheen Kanthawala: Conceptualization, investigation, writing – original draft. Eun Young Choi: Data curation, formal analysis, methodology, writing – original draft. Young Sun Kim: Supervision, Funding acquisition, resources, writing – review & editing.

Acknowledgements
Data was from the Health and Retirement Study, which was funded by the National Institute on Aging (NIA U01AG009740) and the Social Security Administration. The current study was not preregistered. Data, analytic methods, and study materials will be made available upon request through the corresponding author.

Funding
This work was supported by the Ministry of Education of the Republic of Korea and the National Research Foundation of Korea (NRF-2018S1A3A2075402).

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