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Correlates of screen-based behaviors among adults from the 2019 Brazilian National Health Survey

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Abstract

We aimed to investigate correlates of TV viewing and other types of screen-based behaviors in a nationally representative sample of Brazilian adults. In the 2019 Brazilian National Health Survey (including 88,509 adults), TV viewing time and other types of screen behaviors (computer, tablet, and cellphone use) were self-reported and different geographical, sociodemographic, behavioral, and health status factors were investigated as potential correlates. Multinomial logistic regression models were used for the main analyses. Living in capital cities, urban areas, being unemployed, high consumption of soft drinks, obesity, and elevated depressive symptoms were each associated with more TV viewing and more time using other types of screens. There were differential associations between TV viewing and the use of other types of screen across age and socioeconomic variables. For instance, younger adults have a more diverse portfolio of screen time than older adults. To conclude, levels of screen-based behaviors vary by geographical, sociodemographic, behavioral, and health status characteristics. Interventions should focus on high-risk population groups and may benefit from targeting specific sedentary behaviors of interest.

Keywords: Sedentary behavior, Screen time, Television viewing, Adults, Brazil

Introduction

Sedentary behavior represents a considerable public health challenge [1]. Excessive sedentary time increases the risk of morbidity and mortality [2, 3], especially in individuals who are concurrently physically inactive [4], which represents 27.5% of the population [5]. National and international organizations have published guidelines and public health policies which recommend reductions in sedentary behavior, and that endorse population-level surveillance systems [6–9]. Regardless, a high prevalence of sedentary behaviors persists in different parts of

the world, and relative to research conducted within the physical activity realm, surveys on sedentary behaviors at the population level are limited by lack of coverage, being available especially from high-income countries [10, 11].

By definition [12], sedentary behaviors can happen in a range of contexts and domains of life (e.g. reading, writing, using electronic devices), and different behaviors may have a unique profile of correlates and each may be differently associated with health [13]. National health surveys have usually assessed sedentary behavior using crude methods. For example, they have captured limited information about sedentary behaviors as part of questionnaires that have otherwise focused on physical activity (e.g. surveys have captured sitting time by the International Physical Activity Questionnaire [IPAQ] or the Global Physical Activity Questionnaire [GPAQ]), or they have focused only on screen time, especially

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television (TV) viewing. For these reasons, most of the available evidence about the prevalence, temporal trends, and correlates of sedentary behaviors are about total sitting time and TV viewing [10, 14–16].

These concepts are particularly important considering that there is evidence of an emerging "screen transition", with individuals gradually spending less time watching TV and more time engaged with other screens [17]. Providing an enhanced understanding about the detailed correlates of TV viewing, and simultaneously the detailed correlates of other types of screens, could help to assist the development of effective interventions that can be used to limit increases, or target reductions, in specific screen-based behaviors.

To date, studies that have investigated sedentary behavior correlates have consistently found that individual characteristics, including sex, age, illness, other lifestyle behaviors (e.g. tobacco smoking, diet), and sociodemographic factors (e.g. education) are associated with the amount of leisure-time sedentary behavior [18–20]. On the other hand, whilst there is some evidence about the correlates of recreational screen-based behaviors, especially TV viewing, few national surveys have evaluated other types of screen-based behaviors, such as the use of computers, tablets, and cellphones. We aimed to investigate the correlates of different types of screen-based behaviors in a national sample of Brazilian adults.

Methods

Design and sample

This study used data from the 2019 Brazilian National Health Survey, a cross-sectional study conducted with a nationally representative sample of adults in Brazil. The sampling process occurred through three stages. First, census tracts were randomly selected; second, households were randomly selected; third, in each household one inhabitant (aged ≥15y old) was randomly selected. From 100,541 selected households, interviews were conducted with a total of 94,114 participants. In this complete-case analysis, participants aged < 18 years and those with missing data for variables of interest were excluded. The final study sample was composed of 88,509 adults. Further methodological details about the survey are available elsewhere [21]. The Brazilian Council of Ethics in Research approved all procedures according to the Helsinki declaration.

Screen-based behaviors

Participants self-reported the daily time they usually spent engaged in screen-based behaviors. TV viewing was assessed using the question: "How many hours a day do you usually spend watching TV?". The other types of screen-based behavior were assessed through

the question: "How many hours in your daily free time do you usually use a computer, tablet, or cellphone, to access social media, news, videos, games, etc?". For both questions the possible responses were: a) less than 1 h/d; b) more than 1 h/d but less than 2 h/d; c) more than 2 h/d but less than 3 h/d; d) more than 3 h/d but less than 6 h/d; e) more than 6h/d; and f) do not use. For analysis purposes screen time was reclassified into four categories: None (0h/d) / Typical (>0 to <3h/d) / Moderate (≥ 3 to $<6 \, h/d$) / High ($\ge 6 \, h/d$). This classification was based on the Canadian 24h Movement Guidelines [8], which suggests < 3h of recreational screen time per day for adults. In addition, we created a further category of no use, given the specific profiles of non-users of TV-viewing [22], and another category of excessive behavior (twice the recommendation) for sensitivity analyses.

Correlates

A range of correlates across four dimensions was investigated. Possible geographical correlates included the macro regions of the country (North, Northeast, Southeast, South, and Midwest), type of city (capital or non-capital of the State), and characteristics of the surrounding area (urban or rural). Demographic factors included sex (men/women), age group (18-34y, 35-49y, 50-64y, and \geq 65y), highest academic attainment (up to high school, high school, and college or more), income (equivalent to \leq 1 times the minimum wage, 1–3 times the minimum wage, and > 3 times the minimum wage), and internet access (yes/no). The sociodemographic variables were classified based on previous studies from the Brazilian National Health Survey [23, 24].

We also investigated other lifestyle behaviors as correlates, including self-reported leisure-time physical activity (<150 min/week versus \geq 150 min/week, according to the current guidelines from the World Health Organization [6]), consumption of sugary foods and soft drinks (both classified as <5 days/week versus \geq 5 days/week, based on the frequent consumption [25]). Finally, health status factors, comprising elevated depressive symptoms (defined as a score > 9 in the Patient Health Questionnaire-9 [26]), obesity (defined as a body mass index \geq 30 kg/m² [27]), and self-rated health (good/bad [28]) were also investigated as screen time correlates.

Statistical procedures

We used percentages values and 95% confidence intervals to describe the distribution of each group of correlates according to screen time categories. For the main analyses, multinomial logistic regression models were created. Models included each of the groups of geographical, demographic, other lifestyle behaviors, and health status correlates in turn. Variables were tested at each level and

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only individual correlates that were statistically significantly associated with screen time (*p*-value < 0.05) were included in a final model. Estimates were weighted considering the characteristics of the general population as well as the non-response rate [21]. All analyses were conducted in Stata 15, considering sample weights (survey command).

Results

The characteristics of the final sample are presented in Table 1. Of the final sample, 53.2% were women. The weighted prevalence of high (\geq 6.0 h/d) TV viewing and other types of screen-based behavior were 5.8 and 8.6%, respectively. While 8.6% reported no TV viewing, 27.3% reported no use of other types of screen-based behavior. The unadjusted prevalence of TV viewing and other types of screen-based behavior are presented in Supplementary Tables 1 and 2, respectively.

Results of the adjusted multinomial logistic regression models, which were used to identify the main correlates of both TV viewing and other screen-based behaviors, are shown in Tables 2 and 3, respectively. Table 2 shows that higher odds of moderate and high TV viewing were associated with living in the Southeast macro region (compared to the North), being aged ≥65y, unemployment, high soft drink consumption, and obesity. In addition, higher odds of high TV viewing was associated with living in the Northeast macro region (compared to North), and having depressive symptoms. Conversely, lower odds of moderate TV viewing was associated with living in the Midwest macro region (compared to North) and having no internet access, and lower odds of both moderate and high TV viewing was associated with living in non-capitals cities and rural areas, being aged 35-49y, and having higher educational attainment. Factors associated with higher likelihood of not watching TV included living in non-capital cities, being a woman, having the highest level of academic attainment, no internet access, and depressive symptoms. Older age and higher incomes were associated with lower odds of watching no TV.

Table 3 shows that higher likelihood of moderate and high use of other screen types was associated with intermediate educational attainment (compared to the lowest educational status), unemployment, incomes 1–3 times higher than the minimum wage, high consumption of sugary foods, high intake of soft drinks, depressive symptoms, and obesity. In addition, a higher likelihood of moderate use of other screens was associated with the highest level of academic attainment, whereas participants with good self-rated health displayed lower odds of moderate use. In contrast, lower odds of moderate and high use of other types of screen-based behaviors was

Table 1 Characteristics of the sample (n = 88,509)

'		
	%	95%CI
Geographical factors		
Region		
North	7.9	7.7-8.1
Northeast	26.4	26.1-26.9
Southeast	43.4	42.8-44.1
South	14.7	14.3-15.0
Midwest	7.6	7.4-7.8
Type of city		
Capital	41.6	41.1-42.2
Others	58.4	57.8-58.9
Type of residence		
Urban	86.2	85.9-86.5
Rural	13.8	13.5-14.1
Demographic factors		
Sex		
Men	46.8	46.3-47.4
Women	53.2	52.6-53.7
Age group		
18-34y	32.3	31.4-32.5
35-49y	29.3	28.8-29.8
50-64y	22.6	22.1-23.1
≥ 65y	16.1	15.7-16.5
Highest academic achievement		
Up to high school	49.2	48.7-30.4
High school	29.8	29.2-30.4
College or more	21.0	20.5-21.4
Employment status		
Employed	56.2	55.6-56.8
Unemployed	43.8	43.2-44.4
Income		
≤ 1 times minimum wage	51.2	50.6-51.8
1–3 times minimum wage	37.3	36.7-37.9
> 3 times minimum wage	11.5	11.2-11.9
Internet access		
Yes	84.6	84.2-84.9
No	15.4	15.1-15.8
Lifestyle behaviors		
Leisure physical activity		
< 150 min/week	73.5	73.0-74.0
≥ 150 min/week	26.5	26.0-27.0
Sugary foods consumption		
< 5 days/week	85.2	84.7-85.6
≥ 5 days/week	14.8	14.4-15.3
Soft drink consumption		
< 5 days/week	90.8	90.4-91.1
≥ 5 days/week	9.2	8.9-9.6
Health status		
Elevated depressive symptoms		
No	89.2	88.8-89.5

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Table 1 (continued)

	%	95%CI
Yes	10.8	10.5–11.2
Obesity		
No	78.0	77.4–78.5
Yes	22.0	21.5-22.6
Self-rated health		
Good	66.1	65.6-66.6
Bad	33.9	33.4-34.4
Outcomes		
TV viewing		
None (0 h/d)	8.6	8.3-9.0
Typical (> 0 to $< 3 \text{ h/d}$)	69.6	69.0-70.1
Moderate (\geq 3.0 to < 6 h/d)	15.9	15.5-16.4
High (≥6.0 h/d)	5.8	5.6-6.1
Other screen-based behaviors ^a		
None (0 h/d)	27.3	26.8-27.7
Typical (> 0 to < 3 h/d)	50.6	50.0-51.1
Moderate (\geq 3.0 to < 6 h/d)	13.6	13.1-14.0
High (≥6.0 h/d)	8.6	8.3-9.0

Elevated depressive symptoms are defined as a score > 9 in the Patient Health Questionnaire-9. Obesity is defined as a body mass index \ge 30 kg/m²

associated with living in the South macro region (compared to the North), non-capital cities, and rural areas, older age, having no internet access, and physical activity. Except for regional differences, the same factors were associated with higher likelihood of not using other screen types, and the association was strongest for adults aged ≥65y. Other factors associated with higher likelihood of not using other screen types were unemployment, depressive symptoms, and bad self-rated health. Factors associated with lower odds of not using other screen types included living in all macro regions compared to North, being a woman, having higher academic attainment, higher income, and obesity.

Discussion

This investigation shows that living in capital cities, urban areas, being unemployed, poor dietary behaviors, obesity, and elevated depressive symptoms were consistently associated with higher screen time, regardless of type. There were differential associations between TV viewing and use of other types of screen-based behavior across age and socioeconomic variables. Younger adults have a more diverse portfolio of screen time than older adults.

Regarding TV viewing, our results concur with existing evidence, which has shown that higher TV viewing

is correlated with lower educational status, poorer dietary habits, and negative health outcomes [29-31]. The geographical distribution of TV viewing has seldom been studied. We observed that Brazilian adults living in the Northeast and Southeast macro regions were more likely to spend excess time watching TV. In addition, we found that adults living in non-capital cities and rural areas were less likely to watch higher volumes of TV and were also less likely to engage in higher usage of other types of screen-based behavior; in general, they were more likely not to use any screens at all. These results may be related to limited internet availability in rural locations, indeed we found that having no internet access was markedly associated with higher odds of not using other types of screenbased behavior. This pattern of results may also be explained by greater availability and accessibility of public parks and green spaces, lower violence and fewer issues of perceived safety in smaller non-capital cities and rural landscapes [32].

We also identified that unemployment was associated with higher likelihood of moderate and high TV viewing. A u-shaped association was apparent for other types of screen-based behavior, such that unemployment was associated both with higher likelihood of moderate and high use of other types of screen-based behavior, and also with higher odds of not using any other types of screen-based behavior at all. This may be explained by unemployed younger adults using diverse types of screen device throughout the day, and older retirees not using other types of screen. Taken together, the results highlight priority groups that may benefit most from interventions that are designed to reduce screen time.

The main novelty of this study was that we were able to contrast the correlates of TV viewing (which is an important sedentary behavior in its own right and an often used, albeit inadequate, proxy for total sedentary time) with that of other screen types that are increasingly prevalent worldwide [33]. We observed different associations for TV viewing and other screen types across age and socioeconomic variables. For instance, higher educational attainment was associated with lower odds of moderate and high TV viewing, but with higher odds of moderate and high use of all other screens. In addition, the oldest group of adults was more likely to watch moderate and high volumes of TV, but, compared to the youngest age group, older adults of all other ages were less likely to engage in moderate or high usage of other screens, and were more likely to report not using any other screens at all. This pattern of results may reflect social, cultural, and economic differences in the accessibility and usability of different types of screen-based device. The results

CI confidence interval

^a Computer, tablet, or cellphone use to access social media, news, videos, games, etc.

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Table 2 Adjusted regression model quantifying the correlates of TV viewing in Brazilian adults

	TV viewing				
	None (0 h/d) % (95%CI)	Typical (> 0 to < 3 h/d) % (95%CI)	Moderate (≥3.0 to < 6 h/d) % (95%CI)	High (≥6.0 h/d) % (95%CI)	
Geographical factors					
Region					
North	1	Ref	1	1	
Northeast	0.90 (0.82-1.00)	Ref	1.05 (0.96-1.14)	1.32 (1.15–1.50)	
Southeast	0.86 (0.76-0.98)	Ref	1.17 (1.06–1.29)	1.50 (1.31–1.73)	
South	0.74 (0.65-0.85)	Ref	0.99 (0.89-1.11)	0.97 (0.81-1.16)	
Midwest	1.05 (0.92–1.20)	Ref	0.87 (0.77-0.97)	0.99 (0.82-1.18)	
Type of city					
Capital	1	Ref	1	1	
Others	1.22 (1.11-1.34)	Ref	0.82 (0.76-0.88)	0.55 (0.49-0.62)	
Type of residence					
Urban	1	Ref	1	1	
Rural	0.92 (0.83-1.01)	Ref	0.75 (0.69-0.82)	0.56 (0.48-0.65)	
Demographic factors					
Sex					
Men	1	Ref	1	1	
Women	1.11 (1.01–1.23)	Ref	1.00 (0.93-1.07)	1.00 (0.89-1.12)	
Age group	, ,		, ,	, ,	
18-34y	1	Ref	1	1	
35-49y	0.62 (0.55-0.69)	Ref	0.89 (0.80-0.97)	0.76 (0.65-0.89)	
50-64y	0.54 (0.48-0.61)	Ref	0.99 (0.91–1.10)	0.90 (0.77–1.05)	
≥ 65y	0.68 (0.60-0.78)	Ref	1.34 (1.20–1.50)	1.21 (1.03–1.43)	
Highest academic achievement	0.00 (0.00 0.11 0)		115 1 (1125 1155)	(
Up to high school	1		1	1	
High school	0.98 (0.87–1.10)	Ref	1.09 (1.00–1.19)	0.81 (0.71–0.92)	
College or more	1.24 (1.08–1.43)	Ref	0.69 (0.62–0.78)	0.44 (0.36-0.53)	
Employment status	1.24 (1.00 1.43)	nei	0.05 (0.02 0.70)	0.44 (0.50 0.55)	
Employed	1	Ref	1	1	
Unemployed	1.07 (0.97–1.18)	Ref	1.58 (1.46–1.70)	3.02 (2.64–3.45)	
Income	1.07 (0.37–1.10)	IICI	1.50 (1.40-1.70)	3.02 (2.04-3.43)	
	1	Ref	1	1	
≤1 times minimum wage	0.80 (0.71–0.89)	Ref	1.02 (0.95–1.12)	1.00 (0.87–1.14)	
1–3 times minimum wage > 3 times minimum wage	0.68 (0.57–0.82)	Ref	0.97 (0.85–1.12)	0.98 (0.81–1.19)	
g .	0.00 (0.37-0.62)	rei	0.97 (0.65–1.10)	0.96 (0.61–1.19)	
Internet access	1	D-f	1	1	
Yes	1 25 (1 22 1 40)	Ref	1	1	
No	1.35 (1.22–1.48)	Ref	0.91 (0.84–0.99)	1.04 (0.92–1.18)	
Lifestyle behaviors					
Soft drink consumption		0.6			
< 5 days/week	1	Ref	1	1	
≥ 5 days/week	1.03 (0.87–1.22)	Ref	1.31 (1.16–1.48)	1.55 (1.31–1.85)	
Health status					
Elevated depressive symptoms		D (,	
No	1	Ref	1	1	
Yes	1.81 (1.59–2.07)	Ref	1.00 (0.90–1.11)	1.67 (1.45–1.93)	
Obesity					
No	1	Ref	1	1	
Yes	0.96 (0.85–1.09)	Ref	1.24 (1.14–1.32)	1.46 (1.31–1.64)	

The final model is adjusted for all variables presented. Variable with p > 0.05 was removed from the final model. The data are odds ratios and indicate that, for example, compared to participants in urban areas those in rural areas were 44% less likely to be in the high TV viewing than the typical TV viewing group

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Table 3 Adjusted regression model quantifying the correlates of screen time (except TV viewing) in Brazilian adults

	Computer, tablet, or cellphone use to access social media, news, videos, games, etc.			
	None (0 h/d) % (95%CI)	Typical (> 0 to < 3 h/d) % (95%CI)	Moderate (≥3.0 to <6 h/d) % (95%CI)	High (≥6.0 h/d) % (95%Cl)
Geographical factors				
Region				
North	1	Ref	1	1
Northeast	0.88 (0.81-0.96)	Ref	0.99 (0.89-1.10)	1.14 (1.00-1.30)
Southeast	0.56 (0.51-0.62)	Ref	0.92 (0.82-1.04)	0.91 (0.79–1.05)
South	0.58 (0.52-0.65)	Ref	0.79 (0.69-0.89)	0.78 (0.66-0.92)
Midwest	0.55 (0.49-0.63)	Ref	0.94 (0.83-1.07)	0.90 (0.76-1.06)
Type of city				
Capital	1	Ref	1	1
Others	1.25 (1.16-1.35)	Ref	0.87 (0.80-0.95)	0.69 (0.62-0.76)
Type of residence				
Urban	1	Ref	1	1
Rural	1.94 (1.79-2.09)	Ref	0.76 (0.67-0.86)	0.54 (0.45-0.63)
Demographic factors				
Sex				
Men	1	Ref	1	1
Women	0.71 (0.66-0.77)	Ref	1.01 (0.93-1.10)	0.93 (0.84-1.03)
Age group				
18-34y	1	Ref	1	1
35-49y	2.03 (1.84-2.24)	Ref	0.41 (0.37-0.45)	0.27 (0.24-0.30)
50-64y	4.54 (4.10-5.03)	Ref	0.25 (0.23-0.29)	0.14 (0.12-0.16)
≥ 65y	12.61 (11.20–14.19)	Ref	0.21 (0.18-0.26)	0.11 (0.08-0.14)
Highest academic achievement				
Up to high school	1	Ref	1	1
High school	0.44 (0.40-0.47)	Ref	1.37 (1.24–1.51)	1.20 (1.06–1.35)
College or more	0.22 (0.19–0.25)	Ref	1.26 (1.12–1.43)	1.01 (0.87–1.16)
Employment status	, ,		, ,	
Employed	1	Ref	1	1
Unemployed	1.55 (1.43–1.67)	Ref	1.18 (1.08–1.32)	1.74 (1.56–1.95)
Income	,		,	, , , , , , , , , , , , , , , , , , , ,
≤1 times minimum wage	1	Ref	1	1
1–3 times minimum wage	0.64 (0.59-0.69)	Ref	1.19 (1.07–1.31)	1.14 (1.01–1.28)
> 3 times minimum wage	0.48 (0.42-0.56)	Ref	1.10 (0.96–1.28)	1.09 (0.90–1.31)
Internet access	, ,		, ,	,
Yes	1	Ref	1	1
No	5.90 (5.45-6.40)	Ref	0.60 (0.50-0.71)	0.54 (0.43-0.66)
Lifestyle behaviors	,		,	,
Leisure physical activity				
< 150 min/week	1	Ref	1	1
> 150 min/week	1.54 (1.42–1.67)	Ref	0.88 (0.80-0.96)	0.84 (0.76-0.94)
Sugary foods consumption	,	-	,	· · · · · · · · · · · · · · · · · · ·
< 5 days/week	1	Ref	1	1
≥ 5 days/week	0.98 (0.89–1.09)	Ref	1.19 (1.06–1.35)	1.44 (1.28–1.63)
Soft drink consumption		-	((
< 5 days/week	1	Ref	1	1
≥ 5 days/week	1.12 (0.97–1.29)	Ref	1.40 (1.22–1.60)	2.06 (1.79–2.37)

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Table 3 (continued)

	Computer, tablet, or ce	Computer, tablet, or cellphone use to access social media, news, videos, games, etc.			
	None (0 h/d) % (95%CI)	Typical (> 0 to < 3 h/d) % (95%CI)	Moderate (≥3.0 to <6 h/d) % (95%CI)	High (≥6.0 h/d) % (95%CI)	
Health status					
Elevated depressive symptom	oms				
No	1	Ref	1	1	
Yes	1.13 (1.01–1.26)	Ref	1.34 (1.10-1.63)	1.55 (1.34–1.80)	
Obesity					
No	1	Ref	1	1	
Yes	0.90 (0.83-0.97)	Ref	1.19 (1.08-1.32)	1.16 (1.04–1.30)	
Self-rated health					
Good	1	Ref	1	1	
Bad	1.25 (1.16–1.34)	Ref	0.88 (0.80-0.97)	1.09 (0.97-1.24)	

The final model is adjusted for all variables presented. Variable with p > 0.05 was removed from the final model. The data are odds ratios and indicate that, for example, compared to participants without elevated depressive symptoms those with elevated depressive symptoms were 55% less likely to be in the high screen time than the typical screen time group

corroborate previous studies and provide additional support for the concept of an emerging "screen transition", which has previously been shown to be influenced by socioeconomic factors and technological shifts in lowand middle-income countries [17, 31, 34, 35].

It is conceivable that sedentary behaviors may be associated with higher adiposity and adverse health not because of sedentary time per se, but because of coexisting (possibly mediating) obesogenic diet, inactivity, and sleep behaviors [24, 36]. Accordingly, we found that a higher intake of soft drinks was associated with higher TV viewing and higher use of other types of screen-based behavior. Higher consumption of sugary foods was further associated with higher use of other types of screens. The coexistence of TV viewing with poor dietary habits is well established and our findings add to emerging evidencebase for other types of screen-based behavior [37, 38]. We also found being more physically active was associated with lower odds of moderate and high use of other screen types, and with higher odds of not using any other screen types at all.

Regarding health status, obesity and elevated depressive symptoms were associated with higher volumes of TV viewing and also with higher use of other types of screen-based behavior. There was a difference in the shape of associations, however, with some evidence that the associations for depressive symptoms were curvilinear (depressive symptoms were also associated with higher likelihood of not watching TV and not using other screen types). Unfortunately, because this is a cross-sectional study, it is impossible to assign any direction of association to our results. This is particularly problematic for obesity and depressive symptoms, which may exhibit

bidirectional associations with sedentary behaviors [39, 40]. Obesity and depressive symptoms may be both a cause and a consequence of higher TV time and higher use of other screens.

As far as we are aware, this is the first study to explore correlates of different types of screen-based behaviors in a representative sample of Brazilian adults. It is advantageous that we investigated TV viewing and other screens, as well as myriad potential correlates that spanned diverse dimensions. By doing so we have provided an enhanced understanding about the distributions of screen use throughout the Brazilian adult population. This information can be used to assist the development of targeted and more effective interventions to reduce screen-based behaviors in high-risk population groups, such as health education, counseling, household environmental changes and using strategies of usage control by the own devices [41, 42]. A limitation of the current study includes the joint analysis of multiple types of screen-based behavior (computer, tablet, and cellphones) which precluded a more refined analysis of specific screen-based behaviors. It is also a weakness that all data were selfreported, meaning they are subject to inaccurate and biased responses.

To conclude, TV viewing and other screen types share certain correlates among Brazilian adults, however, there are unique screen-type specific correlates. The main differences among the screen types were observed for socioeconomic status and age groups.

Abbreviations

TV: Television; IPAQ: International Physical Activity Questionnaire; GPAQ: Global Physical Activity Questionnaire.

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Supplementary Information

The online version contains supplementary material available at https://doi.org/10.1186/s12889-021-12340-0.

Additional file 1.

Acknowledgements

PC is funded by a BHF Immediate Postdoctoral Basic Science Research Fellowship (FS/17/37/32937). RHOA is supported by the Coordination for the Improvement of Higher Education Personnel (CAPES) with a PhD scholarship (CAPES process: 88887.605034/2021-00). CLS is funded by the National Council for Scientific and Technological Development (Process number 310318/2020-3). AOW is supported by the São Paulo Research Foundation (FAPESP) with a PhD scholarship (FAPESP process: 2019/24124-7). This paper presents independent research. The views expressed in this publication are those of the authors and not necessarily those of the acknowledged institutions.

Authors' contributions

DRS, RHOA, LLB, AOW, and CLS: study design and conception. DRS, PC and AOW: dataset organization and data analysis. DRS: first draft. All authors made substantial contributions on the main manuscript text and approved the final version.

Funding

This specific study did not receive funding. The Brazilian Ministry of Health funds the Brazilian Health Survey.

Availability of data and materials

Data is available under open access through the link: https://www.ibge.gov.br/estatisticas/sociais/saude/9160-pesquisa-nacional-de-saude?=&t=microdados. Accession number: PNS 2019 (*Arquivos de Microdados da PNS 2019*)

Declarations

Ethics approval and consent to participate

The Brazilian Council of Ethics in Research approved all procedures according to the Helsinki declaration and the participants assigned an informed consent form prior to the study beginning.

Consent for publication

The study do not present individual details/images/videos of the participants.

Competing interests

Authors declare no competing interests.

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Received: 11 August 2021 Accepted: 29 November 2021 Published online: 15 December 2021

References

 Owen N, Healy GN, Dempsey PC, Salmon J, Timperio A, Clark BK, et al. Sedentary behavior and public health: integrating the evidence and identifying potential solutions. Annu Rev Public Health. 2020;41:265–87.

- Patterson R, McNamara E, Tainio M, de Sá TH, Smith AD, Sharp SJ, et al. Sedentary behaviour and risk of all-cause, cardiovascular and cancer mortality, and incident type 2 diabetes: a systematic review and dose response meta-analysis. Eur J Epidemiol. 2018;33(9):811–29.
- Saunders TJ, McIsaac T, Douillette K, Gaulton N, Hunter S, Rhodes RE, et al. Sedentary behaviour and health in adults: an overview of systematic reviews. Appl Physiol Nutr Metab. 2020;45(10 (Suppl. 2)):S197–s217.
- Ekelund U, Tarp J, Fagerland MW, Johannessen JS, Hansen BH, Jefferis BJ, et al. Joint associations of accelero-meter measured physical activity and sedentary time with all-cause mortality: a harmonised meta-analysis in more than 44 000 middle-aged and older individuals. Br J Sports Med. 2020;54(24):1499–506.
- Guthold R, Stevens GA, Riley LM, Bull FC. Worldwide trends in insufficient physical activity from 2001 to 2016: a pooled analysis of 358 population-based surveys with 1.9 million participants. Lancet Glob Health. 2018;6(10):e1077–86.
- Bull FC, Al-Ansari SS, Biddle S, Borodulin K, Buman MP, Cardon G, et al. World Health Organization 2020 guidelines on physical activity and sedentary behaviour. Br J Sports Med. 2020;54(24):1451.
- Dempsey PC, Biddle SJH, Buman MP, Chastin S, Ekelund U, Friedenreich CM, et al. New global guidelines on sedentary behaviour and health for adults: broadening the behavioural targets. Int J Behav Nutr Phys Act. 2020;17(1):151.
- Ross R, Chaput JP, Giangregorio LM, Janssen I, Saunders TJ, Kho ME, et al. Canadian 24-hour movement guidelines for adults aged 18-64 years and adults aged 65 years or older: an integration of physical activity, sedentary behaviour, and sleep. Appl Physiol Nutr Metab. 2020;45(10 (Suppl. 2)):557–s102
- Katzmarzyk PT, Powell KE, Jakicic JM, Troiano RP, Piercy K, Tennant B. Sedentary behavior and health: update from the 2018 physical activity guidelines advisory committee. Med Sci Sports Exerc. 2019;51(6):1227–41.
- McLaughlin M, Atkin AJ, Starr L, Hall A, Wolfenden L, Sutherland R, et al. Worldwide surveillance of self-reported sitting time: a scoping review. Int J Behav Nutr Phys Act. 2020;17(1):111.
- 11. Prince SA, Butler GP, Roberts KC, Lapointe P, MacKenzie AM, Colley RC, et al. Developing content for national population health surveys: an example using a newly developed sedentary behaviour module. Arch Public Health. 2019;77:53.
- Tremblay MS, Aubert S, Barnes JD, Saunders TJ, Carson V, Latimer-Cheung AE, et al. Sedentary behavior research network (SBRN) - terminology consensus project process and outcome. Int J Behav Nutr Phys Act. 2017;14(1):75.
- Hallgren M, Dunstan DW, Owen N. Passive versus mentally active sedentary behaviors and depression. Exerc Sport Sci Rev. 2020;48(1):20–7.
- Troiano RP, Stamatakis E, Bull FC. How can global physical activity surveillance adapt to evolving physical activity guidelines? Needs, challenges and future directions. Br J Sports Med. 2020;54(24):1468.
- Prince SA, Melvin A, Roberts KC, Butler GP, Thompson W. Sedentary behaviour surveillance in Canada: trends, challenges and lessons learned. Int J Behav Nutr Phys Act. 2020;17(1):34.
- Prince SA, LeBlanc AG, Colley RC, Saunders TJ. Measurement of sedentary behaviour in population health surveys: a review and recommendations. PeerJ. 2017;5:e4130.
- Mielke GI, Hallal PC, Malta DC, Lee IM. Time trends of physical activity and television viewing time in Brazil: 2006-2012. Int J Behav Nutr Phys Act. 2014;11:101.
- 18. O'Donoghue G, Perchoux C, Mensah K, Lakerveld J, van der Ploeg H, Bernaards C, et al. A systematic review of correlates of sedentary behaviour in adults aged 18-65 years: a socio-ecological approach. BMC Public Health. 2016;16:163.
- Nicolson G, Hayes C, Darker C. Examining total and domain-specific sedentary behaviour using the socio-ecological model - a cross-sectional study of Irish adults. BMC Public Health. 2019;19(1):1155.
- Bakker EA, Hopman MTE, Lee D-c, Verbeek ALM, Thijssen DHJ, Eijsvogels TMH. Correlates of total and domain-specific sedentary behavior: a crosssectional study in Dutch adults. BMC Public Health. 2020;20(1):220.
- Stopa SR, Szwarcwald CL, Oliveira MM, Gouvea ECDP, Vieira MLFP, Freitas MPS, et al. Pesquisa Nacional de Saúde 2019: histórico, métodos e perspectivas. Epidemiol Serv Saúde. 2020;29(5):e2020315.
- 22. Werneck AO, Oyeyemi AL, Szwarcwald CL, Vancampfort D, Silva DR. Associations between TV viewing and depressive symptoms among 60,202

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- Brazilian adults: the Brazilian national health survey. J Affect Disord. 2018:236:23–30.
- Werneck AO, Barboza LL, Araújo RHO, Oyeyemi AL, Damacena GN, Szwarcwald CL, et al. Time trends and sociodemographic inequalities in physical activity and sedentary behaviors among Brazilian adults: National Surveys from 2003 to 2019. J Phys Act Health. 2021;18(11):1332-1341.
- Werneck AO, Oyeyemi AL, Collings PJ, Cyrino ES, Ronque ERV, Szwarcwald CL, et al. Physical activity can attenuate, but not eliminate, the negative relationships of high TV viewing with some chronic diseases: findings from a cohort of 60 202 Brazilian adults. J Public Health (Oxf). 2021;43(1):e7–e15.
- Sousa KT, Marques ES, Levy RB, Azeredo CM. Food consumption and depression among Brazilian adults: results from the Brazilian National Health Survey, 2013. Cad Saude Publica. 2019;36(1):e00245818.
- Santos IS, Tavares BF, Munhoz TN, Almeida LSP, Silva NTB, Tams BD, et al. Sensibilidade e especificidade do Patient Health Questionnaire-9 (PHQ-9) entre adultos da população geral. Cad Saúde Pública. 2013;29:1533–43.
- Purnell JQ. Definitions, classification, and epidemiology of obesity. In: Feingold KR, Anawalt B, Boyce A, Chrousos G, de Herder WW, Dhatariya K, Dungan K, Grossman A, Hershman JM, Hofland J, et al., editors. Endotext. South Dartmouth: MDText.com, Inc. Copyright © 2000–2021, MDText. com, Inc.; 2000.
- Silva DAS. Longer leisure walking time is associated with positive selfrated health among adults and older adults: a Brazilian nationwide study. PeerJ. 2021;9:e11471.
- Werneck AO, Cyrino ES, Collings PJ, Ronque ERV, Szwarcwald CL, Sardinha LB, et al. TV viewing in 60,202 adults from the National Brazilian Health Survey: prevalence, correlates, and associations with chronic diseases. J Phys Act Health. 2018;15(7):510–5.
- Scully M, Dixon H, Wakefield M. Association between commercial television exposure and fast-food consumption among adults. Public Health Nutr. 2009;12(1):105–10.
- 31. Iven Mielke G, Ricardo L, Wendt A, Christofaro D, Owen N. Descriptive epidemiology of high TV-viewing time in Brazilian adults. Rev Bras Atividade Física Saúde. 2018;23:1–6.
- Prince SA, Reed JL, McFetridge C, Tremblay MS, Reid RD. Correlates of sedentary behaviour in adults: a systematic review. Obes Rev. 2017;18(8):915–35.

- Vizcaino M, Buman M, DesRoches T, Wharton C. From TVs to tablets: the relation between device-specific screen time and health-related behaviors and characteristics. BMC Public Health. 2020;20(1):1295.
- 34. Mielke GI, da Silva IC, Owen N, Hallal PC. Brazilian adults' sedentary behaviors by life domain: population-based study. PLoS One. 2014;9(3):e91614.
- Mielke GI, Brown WJ, Ekelund U, Brage S, Goncalves H, Wehrmeister FC, et al. Socioeconomic position and sedentary behavior in Brazilian adolescents: a life-course approach. Prev Med. 2017;107:29–35.
- Werneck AO, Oyeyemi AL, Szwarcwald CL, Stubbs B, Silva DR. Potential influence of physical, psychological and lifestyle factors on the association between television viewing and depressive symptoms: a crosssectional study. Gen Hosp Psychiatry. 2019;60:37–43.
- Werneck AO, Silva DR, Malta DC, Gomes CS, Souza-Júnior PR, Azevedo LO, et al. Associations of sedentary behaviours and incidence of unhealthy diet during the COVID-19 quarantine in Brazil. Public Health Nutr. 2021;24(3):422–6.
- Sisson SB, Shay CM, Broyles ST, Leyva M. Television-viewing time and dietary quality among U.S. children and adults. Am J Prev Med. 2012;43(2):196–200.
- Barone Gibbs B, Aaby D, Siddique J, Reis JP, Sternfeld B, Whitaker K, et al. Bidirectional 10-year associations of accelerometer-measured sedentary behavior and activity categories with weight among middle-aged adults. Int J Obes. 2020;44(3):559–67.
- Vancampfort D, Firth J, Schuch FB, Rosenbaum S, Mugisha J, Hallgren M, et al. Sedentary behavior and physical activity levels in people with schizophrenia, bipolar disorder and major depressive disorder: a global systematic review and meta-analysis. World Psychiatry. 2017;16(3):308–15.
- 41. Wu L, Sun S, He Y, Jiang B. The effect of interventions targeting screen time reduction: a systematic review and meta-analysis. Medicine (Baltimore). 2016;95(27):e4029.
- Shrestha N, Grgic J, Wiesner G, Parker A, Podnar H, Bennie JA, et al. Effectiveness of interventions for reducing non-occupational sedentary behaviour in adults and older adults: a systematic review and metaanalysis. Br J Sports Med. 2019;53(19):1206.

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