



Within- and between-person associations of time-varying determinants with snacking in older adults: an ecological momentary assessment study

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Abstract

Promoting healthy snacking is important in addressing malnutrition, overweight and obesity among an ageing population. However, little is known about the factors underlying snacking behaviour in older adults. The present study aimed to explore within- and between-person associations between determinants (i.e. intention, visibility of snacks, social modelling and emotions) and snacking behaviours (i.e. decision to snack, health factor of the snack and portion size) in older adults (60+). Conducting a two-part intensive longitudinal design, data were analysed from forty-eight healthy older adults consisting of (1) an event-based self-report ecological momentary assessment (EMA) diary every time they had a snack and (2) a time-based EMA questionnaire on their phone five times per day. Analysis through generalised linear mixed models indicated that higher intention to snack healthily leads to healthier snacking while higher levels of social modelling and cheerfulness promote unhealthier choices within individuals. At the between-person level, similar results were found for intention and social modelling. Visibility of a snack increased portion size at both a within- and between-person level, while the intention to eat a healthy snack only increased portion size at the between-person level. No associations were found between the decision to snack and all determinants. This is the first study to investigate both within- and between-person associations between time-varying determinants and snacking in older adults. Such information holds the potential for incorporation into just-in-time adaptive interventions, allowing for personalised tailoring, more effective promotion of healthier snacking behaviours and thus pursuing the challenge of healthy ageing.

Keywords: Ecological momentary assessment: Emotions: Social modelling: Intention: Elderly

Worldwide, both the proportion and the absolute number of older adults are increasing rapidly. It is estimated that the population aged ≥ 60 years will grow from 10% of the total population in 2021 to nearly 17% by 2050^(1,2). This demographic shift necessitates proactive measures to boost healthy ageing, defined by the WHO as ‘the process of developing and maintaining the functional ability that enables wellbeing in older age’⁽³⁾. Lifestyle behaviours, including a healthy diet (e.g. healthy snacks), are particularly important in pursuing the challenge of healthy ageing⁽⁴⁾. Existing literature shows a clear link between snacking behaviour and health status in older adults, which can yield positive and negative health effects^(5–8). Negative effects are linked to a disturbed energy and nutrient balance. The consumption of energy-dense snacks can contribute to higher energy intake and subsequent weight gain⁽⁸⁾. Still, this does not mean that snacking automatically leads to weight gain as this depends on various factors, including frequency, portion size and type of snack. It is important to note, however,

that depending on nutritional requirements, what is considered healthy for some older adults, is not necessarily healthy for all older adults. It has been shown, for example, that the risk of malnutrition, an inadequate intake of energy and nutrients, is higher in older adults compared with other age groups⁽⁵⁾. Reduced appetite is often reported as one of the main factors that lead to a reduction in energy or nutrient intake⁽⁵⁾. Therefore, specifically in older adults suffering from malnutrition, consuming energy-dense snacks between meals may be an effective way to increase energy intake and improve nutritional status⁽⁶⁾. On the contrary, a potential positive effect of snacking is related to sarcopenia (i.e. progressive loss of skeletal muscle mass and strength)⁽⁹⁾. Snacks containing protein help both in preventing and treating sarcopenia and are therefore not only beneficial for malnourished individuals but for all older adults, despite their satiating effect and the risk of reducing intake at the next meal⁽¹⁰⁾. Malnutrition is, however, beyond the scope of this paper and will not be discussed further in detail. Altogether, it is clear that

Abbreviation: EMA, ecological momentary assessment.

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snacking can have both positive and negative consequences on the health of older adults, making it an important health behaviour to investigate further.

The phenomenon of snacking gradually made its way into daily eating patterns over the past decades, particularly in high-income countries⁽⁷⁾. In the present paper, snacking is defined as an eating moment that falls outside of the three main meals (i.e. breakfast, lunch and dinner)⁽¹¹⁾. Only solid foods, but not beverages, were assessed to reduce confusion surrounding snack definitions⁽¹²⁾ and to reduce participant burden⁽¹³⁾. Studies on snacking among older adults reveal diverse prevalence numbers, with variations in the energy contribution of snacks to total daily intake ranging across European countries from 6% in France to 32% in The Netherlands⁽¹⁴⁾. Several studies concluded that older adults have steeper peaks in energy intake during the main meals compared with young adults^(15,16). The proportion of older adults who consumed at least one snack per day ranged from 77 to 84%, which is significantly higher compared with other age groups^(7,15,17).

To achieve a comprehensive understanding of snacking behaviour in older adults, it is important to identify the underlying factors. The present study identifies these determinants by using a socio-ecological model as the guiding theoretical framework⁽¹⁸⁾. This model categorises factors affecting food intake into three levels: macro- (policies), meso- (community settings) and micro-level (inter- and intrapersonal factors). The present study focuses on determinants at the meso- (i.e. visibility of snacks^(19,20)) and micro-level (i.e. emotions^(21,22), intention⁽²³⁾ and social modelling^(19,24)). A considerable amount of literature has been published on the drivers of eating behaviour and some are briefly outlined here. First, literature on the influence of emotions is ambiguous. Previous research has demonstrated that negative emotions negatively affect dietary intake (e.g. among older adults, feelings of loneliness have been linked to malnutrition)⁽²²⁾. In the context of diet quality, a study in older adults showed that mood was not an important barrier or motivator in making food choices⁽²⁵⁾. Conversely, other studies found that positive affect was associated with healthier dietary patterns, while negative affect was not significantly related to dietary quality^(26,27). The paradox of emotional well-being in ageing suggests that older adults often maintain or even improve their emotional well-being despite physical and mental decline^(28,29). This phenomenon is attributed to older adults' superior emotional regulation skills compared with younger adults^(30,31). With this in mind, more research is needed to clarify the role of emotions in dietary behaviour among older adults. Second, according to the socio-ecological model, someone's intention is an important determinant of behaviour. Nonetheless, high levels of intention to eat healthy will not automatically lead to actual healthy eating behaviour. Thus, someone's intention does play a role, but can be influenced by, for example, environmental cues (e.g. seeing others eat or having snacks available)⁽¹⁹⁾. Lastly, Elliston *et al.* found that observing others eat increased the likelihood of eating because being exposed to food increases subjective experiences of desire to eat^(19,20). However, older adults exhibit more habitual behaviours and eating patterns, possibly making them less susceptible to social modelling compared with younger adults⁽³²⁾. Differences in

motives for the consumption of regular meals (i.e. breakfast, lunch and dinner) compared with motives for snack consumption can be identified^(24,33–36). On the one hand, individuals commonly attribute eating motives such as habit and hunger to regular meals⁽³³⁾. On the other hand, for snack consumption, emotional and external motives (e.g. visual and aromatic food cues) are reported more frequently^(33,35). Nevertheless, to our knowledge, there is a notable dearth of studies examining the association between these determinants and snacking behaviour, particularly among older adults.

Previous studies on snacking have typically relied on traditional assessment methods (e.g. dietary recall), but these methods have several limitations including low compliance rates, under-reporting of events⁽³⁷⁾ or backfilling of past events⁽³⁸⁾. Moreover, these methods usually assess determinants only once and at the start of the study, while in reality determinants may vary throughout the day⁽³⁹⁾. To address these limitations, ecological momentary assessment (EMA) emerges as a promising solution, with previous studies already confirming that it is a feasible method to examine eating behaviour in older adults^(13,40). EMA enables the assessments of determinants in real-life situations at multiple moments throughout the day and maximises ecological validity and minimises recall bias⁽⁴¹⁾. These repeated assessments are necessary to reveal underlying dynamic behavioural processes⁽⁴⁰⁾ and offer useful information to develop personalised or just-in-time adaptive interventions, which aim to provide the right type and intensity of support to individuals at the right time⁽⁴²⁾. For example, suggestions to eat a healthy snack can be adjusted to the emotional 'state' of the individual, i.e. when the individual is more prone to engage in unhealthy snacking.

The purpose of the current study was to investigate within- and between-person associations of various determinants at the meso- and micro level of the socio-ecological model and snacking behaviours (i.e. decision to snack, health factor of the snack and portion size) in older adults. The uniqueness of this paper lies mainly in the repeated-measures design throughout the day, which allows to draw conclusions about both between- and within-person associations. Because there has been little research conducted on the determinants of snacking within persons, especially in older adults and with repeated assessments within days, this study is primarily exploratory in nature.

Methods

Procedure and participants

Fifty-one healthy older adults, aged 60 or above, were recruited for this prospective observational study with an intensive longitudinal design. Participants were recruited through convenience and snowball sampling in the personal network of the researchers and via community associations for older adults. Inclusion criteria were (1) no impaired cognition (i.e. diagnosis of dementia; Alzheimer or other cognitive diseases), (2) living at home independently, (3) no impairment of fine motor skills (i.e. being able to complete data collection with a smartphone), (4) no severe impairment of vision and/or hearing (i.e. being able to hear the triggers for the EMA questions and read the





questions on the smartphone) and (5) be Dutch-speaking. During the initial contact via email, the inclusion criteria were discussed with the participant. Data were collected between August and November 2021. This study was conducted according to the guidelines laid down in the Declaration of Helsinki, and all procedures involving human subjects were approved by the Ghent University Hospital Ethics Committee (registration number B6702021000698). Written informed consent was obtained from all subjects.

During a first home visit, all participants completed a baseline questionnaire on socio-demographic variables. Thereafter, face-to-face instructions for the measurement period were given and participants received a short introduction on how to use the smartphone-based EMA application, supported by a user manual (online Supplementary file 1). They were encouraged to use their own smartphone (the lowest operating systems used were Android 5.0 and iOS 12.4), but if this was impossible a smartphone (Motorola Moto G20 64GB, Android 6.0) was provided by the researchers. The first home visit was followed by a measurement period of seven consecutive days (i.e. 5 weekdays and 2 weekend days), during which the participants answered a combination of time-based and event-based EMA questionnaires. Data collection started on a randomly selected day during the week, with this starting day varying between participants. Afterwards, all measurement instruments were collected during a second home visit and a short follow-up interview took place. The study design is presented in Fig. 1.

Questionnaires

Baseline. Participants self-reported the socio-demographic variables age, gender, height, weight, educational level, main occupation (before retirement) and marital status. Waist circumference was measured by the researchers using a tape measure.

Event-based ecological momentary assessment. Participants were asked to keep a paper-based snack diary, where they self-reported specific information every time they consumed a snack (i.e. decision to snack). A snack was defined as a solid food that was consumed outside of the three main meals (i.e. breakfast, lunch and dinner)⁽¹¹⁾. For every snack, participants reported time, the type of snack, portion size (in grams, household measures (e.g. a teaspoon) or by using a food model booklet⁽⁴³⁾), visibility of the snack and social modelling. Visibility was queried with the item ‘Was the snack visible when you decided to snack?’⁽²⁰⁾. Participants were instructed to answer (Yes) if the snack was visible in their immediate environment. They were instructed to answer (No) if the snack was available but not visible or if they bought the snack at the store. Social modelling was questioned with the item ‘While eating your snack, could you see someone else eating in your surroundings?’ (Yes, No)⁽¹⁹⁾.

Thereafter, the UK Nutrient Profiling Model was used by the researchers to calculate a health factor and divide the types of snacks into two categories (i.e. healthy and less healthy)⁽⁴⁴⁾. The scoring system uses a formula taking into account energy content (kJ/100 g), saturated fat (g/100 g), sugar (g/100 g), salt (mg/100 g), fruit-vegetables-nuts (%), fibre (g/100 g) and protein (g/100 g) to classify a snack as ‘healthy’ (< 4) or ‘less healthy’ (> 4). The Dutch Food Composition Database (NEVO online version 2021/7.1) was used to analyse the nutrients of the foods mentioned in the snack diary. When portion sizes were not recorded in weight units, they were converted into grams through the utilisation of the Belgian manual for standardised quantification of food⁽⁴⁵⁾.

Time-based ecological momentary assessment. Prior studies have indicated that a minimum of five daily assessments is necessary to effectively record the dynamic fluctuations of determinants linked to health behaviours within individuals^(46,47).

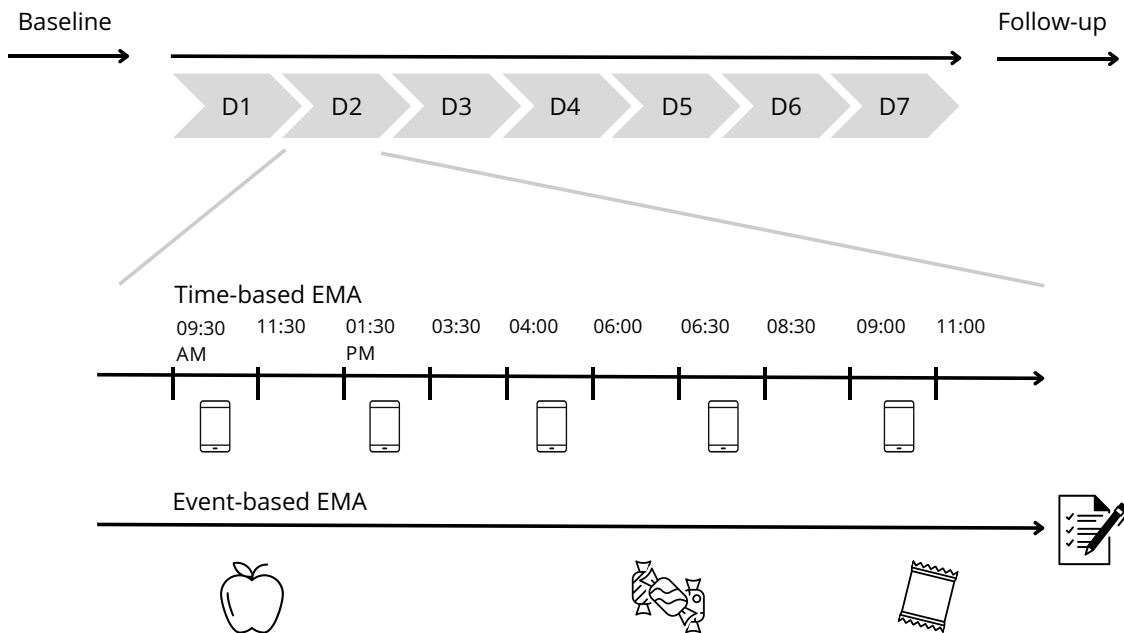


Fig. 1. Example of the study design.

Therefore, participants received auditory signals on their smartphone five times per day for seven consecutive days, alerting them to fill in a questionnaire using the Smartphone Ecological Momentary Assessment³ application (i.e. the participants were asked to answer the questionnaire thirty-five times in total)⁽⁴⁸⁾. Smartphone Ecological Momentary Assessment³ randomly selected a triggering time within 2-hour intervals, being: 09.30–11.30, 13.30–15.30, 16.00–18.00, 18.30–20.30 and 21.00–23.00. The timeframes were chosen based on previous research looking at prevalence of snacking in older adults throughout the day⁽¹⁵⁾. The minimum interval between two successive signals was 30 min. If participants did not immediately respond to the signal, they received up to two reminders. If the questionnaire was still unanswered after 30 min, the questionnaire became inaccessible. The assessment time used in further analysis was the time point at which the questionnaire was completed. In the time-based EMA questionnaire, intention for healthy snacking and emotions were assessed. Intention for healthy snacking was assessed by the item '*If I eat a snack in the next 2 h, I want to choose a healthy snack*' and answers ranged from (*strongly disagree*) to (*strongly agree*) on a seven-point Likert scale⁽⁴⁹⁾. Whether or not a snack was seen as (un)healthy was based on the subjective opinion of the participant. Five negative emotions (i.e. loneliness, insecurity, anxiousness, irritation and feeling down) and four positive emotions (i.e. cheerfulness, relaxation, enthusiasm and satisfaction) were selected from a list of items that are frequently used for EMA by the Department of Psychiatry and Neuropsychology at the University of Maastricht⁽⁵⁰⁾. These nine emotions were captured on a seven-point Likert scale in a similar way. An example of an item is '*How lonely were you just before you received the trigger?*' with answers ranging from (*not at all lonely*) to (*very lonely*).

Statistical analyses

The data were analysed using R version 4.2.3⁽⁵¹⁾. To investigate the effect of the determinants on participants' snacking behaviour, the time-based and event-based datasets were combined. If a snack was consumed in the subsequent 2 h following the time-based EMA trigger, all corresponding details from the snack diary were associated with the respective time-based EMA trigger. In case multiple time-based EMA triggers preceded a snack, the trigger closest to the eating moment was chosen. If multiple snacking moments (i.e. event-based EMA) could be associated with one trigger, both were linked to the same time-based EMA information. The determinants visibility and social modelling were not included in the analyses for the dependent variable 'decision to snack', since they were only assessed if people actually snacked. The determinant health factor was rescored to a positive number by adding ten units to each value to allow further analysis to be interpreted more easily. To take the clustering of the data within participants into account, (generalised) linear mixed models (two-level models, triggers within individuals), as implemented in the package lme4 version 1.1–34, were used to analyse the data⁽⁵²⁾. The percentage of between- and within-subject variance of the determinants was calculated by running intercept-only models (i.e. models only including a fixed and random intercept) with each of the

determinants as outcome variables. To examine the effect of the determinants on participants' snacking behaviour (outcome variables: decision to snack, health factor and portion size), different models were fitted with the determinants as between-subject (i.e. mean of the variable at the subject level) as well as within-subject (i.e. individuals' score at the time of the trigger minus their mean score) variables. For the outcome variable decision to snack (yes; no), logistic regression models were used. Two models (i.e. a random intercept-fixed slope model and a random intercept-random slope model) were fitted for each determinant separately, which led to twenty-four models in total (i.e. twelve determinants × two models). Of those two models per determinant, the model with the lowest Akaike's information criterion value indicating a better model fit was chosen. For the outcome variables health factor and portion size, three different generalised linear models (i.e. Gaussian with identity, Gamma with identity and Gamma with log) and both a random intercept-fixed slope model and random intercept-random slope model were tested for each determinant separately. To check which model fitted the data best, the Akaike's information criterion value was defined. To standardise the analyses, one model was chosen per outcome variable, since the Akaike's information criterion values across models were very similar. The random intercept-fixed slope model with gamma variance and identity function was applied in all analyses for the outcome variable health factor and the random intercept-fixed slope model with γ variance and log function was used for portion size. Model assumptions were visually checked (i.e. normality of the residuals, homoscedasticity, outliers and influential observations). Statistical significance was determined at $P < 0.05$.

Power analysis

Since this is an exploratory study, the sample size was determined aiming to avoid overfitting (i.e. a model becomes more complicated than necessary and is modelling spurious relationships) instead of obtaining the desired power. In general, for linear and logistic regression models, a minimum of ten observations per predictor allows good estimates^(53,54). Because univariable models were used in the current study, leading to only one degree of freedom per model, we set a minimum of ten participants for the models with continuous outcome variables. Previous studies report snacking prevalence ranging from 77 to 84% in older adults^(17,55). Therefore, in the models with a binary outcome measure, assuming that 80% of the participants consume snacks, ten participants account for the 20% in the smallest group who do not snack on a daily basis. Added together with forty participants who do snack ($10/0.2 \times 0.8$), this results in a minimum of fifty participants to be recruited. However, the above reasoning has an important limitation. In EMA studies, the collected data are not independent but hierarchical (two-level models, triggers within individuals). Neglecting the clustering when modelling data frequently leads to underestimated standard error estimates, especially when the outcome variable shows dependence based on the clustering of data. The impact of the number of clusters on model estimates is moderated by the sample size within these clusters. Although a





specific sample size to ensure unbiased estimates cannot be pinpointed, prior research recommends a minimum of thirty participants (representing the number of clusters) with thirty observations per participant, assuming equal sample sizes within each cluster⁽⁵⁶⁾. A review found that the compliance rate for EMA studies in European older adults was on average 86.19%⁽¹³⁾, which confirms the feasibility of this method for data collection among older adults. Therefore, presuming a compliance rate of 86% for older adults in EMA studies, we aimed to include a minimum of fifty participants to which each a total of thirty-five (30/0.86 × 1) triggers were sent.

Results

Descriptive statistics

In total, fifty-one older adults completed the measurements for a period of seven consecutive days. Based on the time-based assessments, three individuals had an overall response rate lower than 33% (i.e. < 12 of 35 triggers) on the EMA questionnaire and were excluded from further analyses, in line with methodology used in other research^(39,57). In a follow-up interview, these drop-out participants reported difficulties with the use of the smartphone or indicated to be unable to answer signals during working hours. Ultimately, forty-eight participants were included in the analyses (smartphone ownership: 91.7%). The characteristics of the study population are presented in [Table 1](#).

Throughout the study, a total of 1680 time-based EMA triggers were sent. Of these triggers, 475 were not answered and six contained missing information, which led to their exclusion from the analysis. Eventually, 1199 triggers were completed (overall response rate = 71.4%; mean of 23.9 completed questionnaires per participant) and were included for analysis of the dependent variable decision to snack. In 35.0% of the moments where a time-based EMA questionnaire was triggered, participants reported a snack moment (event-based EMA) in the subsequent 2 h following the trigger. Out of the 761 snacking moments

Table 2. Descriptive statistics of the snack diary

Item	Total sample n 761 (%)	M	SD	Q1–Q3	Min–max
Snack frequency (snacks/day)		2.13	1.32	1–3	0–6
Time of day					
00.01–06.00	0.0				
06.01–12.00	25.2				
13.01–18.00	41.3				
18.01–00.00	32.9				
Snack type					
Fruit	30.8				
Pastries and cakes	24.7				
Sugar, sweets and sweet toppings	15.5				
Milk(products)	8.9				
Nuts and seeds	4.4				
Salty snacks	2.6				
Other*	13.1				
Portion size (g)		82.2	67.7	30–125	1–425
Health factor					
Healthy†	47.0	7.0	12.7	–2–23	–8–31
Visibility (Yes)	33.9				
Social modelling (Yes)	43.8				

M, mean. SD, standard deviation. Min, minimum. Max, maximum.
* (e.g. cheese, bread, charcuterie, vegetables ...).
† Health factor < 4.

recorded in the snack diaries, 420 could be matched with a time-based EMA trigger and were included for further analysis of the dependent variables health factor and portion size. [Table 2](#) contains the descriptive statistics pertaining to the snack diary. [Online Supplementary file 2](#) provides more detailed information about the snack frequency per participant individually.

[Table 3](#) displays the mean scores, the between-subject variance and the within-subject variance between days and within days of the time-based determinants. For the determinants intention, loneliness, relaxation, enthusiasm, anxiousness and feeling down, the between-subject variance (i.e. ranging from 48.6% to 66.3%) is higher than the within-subject variance within days (i.e. ranging from 31.1% to 48.0%). The opposite is true for the determinants cheerfulness, satisfaction, insecurity and irritation (i.e. the between-subject variance ranges from 30.4% to 47.7% and the within-subject variance within days ranges from 47.9% to 64.4%). Within-subject variance between days explained only 1.5% to 14.6% of the variance, indicating that the variation of the examined determinants within persons, but between days, was limited. [Online Supplementary file 3](#) provides the individual scores on the determinants over the measurement period. This way, an overview of the extent to which the determinants fluctuate is given.

Within- and between-person associations with the decision to snack

[Table 4](#) shows the results for the logistic regression models of the determinants related to the decision to snack (yes/no). No significant within- or between-person associations were found between the determinants and the decision whether or not to snack within 2 hours after the trigger.

Table 1. Descriptive characteristics of the participants

Demographics	Total sample n 48 (%)	M	SD	Min–max
Age (years)		66.2	6.2	60–85
Women	52.9			
BMI (kg/m ²)		25.2	3.6	17.8–33.4
Waist circumference (cm)		97.8	9.8	70–122
Non-tertiary education	33.3			
Main occupation (before retirement)				
Household	3.9			
Blue collar worker*	19.6			
White collar worker†	72.5			
Other	4.0			
Marital status				
Single	3.1			
Married or living together	84.4			
Divorced	3.1			
Widow(er)	9.4			

M, mean; Min, minimum; Max, maximum.
* Self-employed, worker.
† Employee, education, executives, free professions and officer.

Table 3. Mean scores, between subject variance, within-subject variance (between days – within days)

Determinants	M	SD	Q1–Q3	Min–max	Between subject variance (%)	Within-subject variance – between days (%)	Within-subject variance – within days (%)
Intention	4.4	1.5	4–5	1–7	66.3	2.6	31.1
Loneliness	1.5	0.8	1–2	1–7	52.6	3.4	44.0
Cheerfulness	4.6	0.9	4–5	1–7	37.5	14.6	47.9
Relaxation	5.3	1.1	5–6	1–7	48.8	6.5	44.7
Enthusiasm	4.5	1.0	4–5	1–7	48.6	10.1	41.4
Satisfaction	5.1	1.0	4–6	2–7	47.7	1.5	50.9
Insecurity	1.5	0.8	1–2	1–5	45.8	4.8	49.4
Anxiousness	1.4	0.7	1–2	1–7	49.3	2.7	48.0
Irritation	1.6	0.9	1–2	1–6	30.4	5.2	64.4
Feeling down	1.5	0.8	1–2	1–5	56.2	3.5	40.3

M, mean; Min, minimum; Max, maximum.

Table 4. Within- and between-person associations between different possible determinants of snacking with the decision to snack, health factor and portion size

Decision to snack (yes, no)	Within-subject associations			Between-subject associations		
	Logistic regression model			Logistic regression model		
	OR	95 % CI	P-value	OR	95 % CI	P-value
Intention	1.02	0.78, 1.33	0.87	0.91	0.70, 1.19	0.50
Loneliness	1.03	0.81, 1.30	0.82	0.98	0.59, 1.61	0.93
Cheerfulness	1.02	0.85, 1.23	0.80	1.15	0.72, 1.85	0.56
Relaxation	0.99	0.83, 1.17	0.89	1.06	0.72, 1.57	0.77
Enthusiasm	1.03	0.84, 1.25	0.80	1.16	0.76, 1.76	0.50
Satisfaction	0.94	0.78, 1.14	0.55	0.96	0.62, 1.48	0.85
Insecurity	1.01	0.81, 1.25	0.93	1.20	0.72, 2.00	0.48
Anxiousness	0.98	0.74, 1.29	0.88	1.24	0.68, 2.27	0.49
Irritation	0.91	0.76, 1.09	0.29	1.28	0.73, 2.24	0.38
Feeling down	0.90	0.69, 1.18	0.44	1.19	0.72, 1.95	0.50

Health factor	Gamma with identity model			Gamma with identity model		
	B	95 % CI	P-value	B	95 % CI	P-value
Intention	-2.53	-4.17, -0.90	<0.001	-2.81	-4.50, -1.11	<0.001
Visibility	-0.12	-2.89, 2.64	0.93	-3.17	-11.63, 5.29	0.46
Social modelling	6.57	4.13, 9.00	<0.001	7.66	-1.23, 16.55	0.09
Loneliness	-1.46	-3.32, 0.40	0.12	-2.18	-5.95, 1.59	0.26
Cheerfulness	2.14	0.12, 4.16	0.04	0.15	-3.08, 3.38	0.93
Relaxation	-1.21	-3.06, 0.65	0.20	0.61	-2.25, 3.46	0.68
Enthusiasm	1.95	-0.12, 4.02	0.07	0.31	-2.69, 3.31	0.84
Satisfaction	1.13	-1.09, 3.35	0.32	1.61	-1.52, 4.75	0.31
Insecurity	0.74	-1.48, 2.97	0.51	-0.74	-4.45, 2.96	0.69
Anxiousness	-0.29	-2.79, 2.21	0.82	-0.18	-4.81, 4.46	0.94
Irritation	-0.80	-2.46, 0.86	0.35	-2.26	-6.22, 1.71	0.26
Feeling down	1.05	-1.52, 3.61	0.42	-1.68	-4.84, 1.47	0.30

Portion size	Gamma with log model			Gamma with log model		
	ExpB	95 % CI	P-value	ExpB	95 % CI	P-value
Intention	0.99	0.90, 1.09	0.85	1.15	1.02, 1.29	0.03
Visibility	1.03	1.02, 1.04	<0.001	1.35	1.33, 1.36	<0.001
Social modelling	1.13	0.94, 1.35	0.18	0.89	0.53, 1.50	0.66
Loneliness	1.00	0.86, 1.17	0.99	1.09	0.85, 1.39	0.49
Cheerfulness	0.99	0.88, 1.12	0.90	0.93	0.76, 1.14	0.50
Relaxation	1.11	0.98, 1.25	0.11	1.05	0.87, 1.27	0.62
Enthusiasm	1.07	0.92, 1.24	0.41	0.90	0.75, 1.07	0.23
Satisfaction	0.98	0.85, 1.13	0.81	0.90	0.74, 1.09	0.27
Insecurity	1.01	0.88, 1.15	0.88	1.01	0.79, 1.28	0.97
Anxiousness	1.08	0.91, 1.28	0.38	0.95	0.71, 1.27	0.73
Irritation	0.99	0.89, 1.11	0.85	1.11	0.86, 1.43	0.41
Feeling down	0.94	0.79, 1.11	0.47	1.07	0.87, 1.33	0.53

ExpB, exponentiated estimate (representing a proportional difference in portion size). Significant results ($P < 0.05$) are highlighted in bold.

Within- and between-person associations with health factor

Table 4 provides the results for the models with γ variance and identity function of the determinants related to the health factor of the snack. A significant within-person association was found for intention to eat a healthy snack ($P < 0.01$), social modelling ($P < 0.01$) and cheerfulness ($P = 0.04$). If a person's intention to eat a healthy snack is one unit higher than his own average intention level, the health factor score is associated with a decrease of 2.53 units (= healthier snack). If a person sees someone eating in their environment compared with when they do not see someone eating in their environment, the health factor is associated with an increase of 6.57 units (= unhealthier snack). If a person scores one unit higher on cheerfulness than his own average cheerfulness level, the health factor is associated with an increase of 2.14 units (= unhealthier snack). A significant between-person association was found for intention ($P < 0.01$). Per one-unit increase in intention to eat a healthy snack, the health factor is associated with a decrease of 2.81 units (= healthier snack). No significant associations were found for the other determinants.

Within- and between-person associations with portion size

Table 4 also provides the results for the models with γ variance and log function of the determinants related to the portion size of the snack. A significant within-person association was found for visibility ($P < 0.01$) and significant between-person associations were found for intention ($P = 0.03$) and visibility ($P < 0.01$). If a snack is visible to the person compared with when the snack is not visible for that same person, the portion size of the snack is associated with an increase of 3%. Per one-unit increase in intention, the portion size is associated with an increase of 15%. If a snack is visible compared with when it is not visible, the portion size is associated with an increase of 35%. No other significant associations were found.

Discussion

This study examined the within- and between-person associations of multiple determinants on snacking behaviours (i.e. decision to snack, health factor and portion size). At the intra-individual (within-person) level, intention to eat a healthy snack was negatively associated with health factor (= healthier snack), social modelling and cheerfulness were positively associated with health factor (= unhealthier snack), and visibility was positively linked with portion size. Furthermore, at the inter-individual (between-person) level, intention to eat a healthy snack was negatively associated with health factor (= healthier snack), and intention to eat a healthy snack and visibility were positively linked with portion size. However, the other determinants (i.e. loneliness, relaxation, enthusiasm, satisfaction, insecurity, anxiousness, irritation, feeling down) were not related to health factor and portion size, nor at the within-person, nor the between-person level. Strikingly, none of the included determinants were related to the decision of whether or not to eat a snack at the within- or between-person level. To our

knowledge, this is the first study focusing on within-person associations of specific determinants with snacking in an older age group, so most findings can only be compared with between-person associations examined in other studies.

The intention to eat a healthy snack leads to actually eating a healthier snack and this applies to both the within- and between-person level. To put it another way, both an increase in intention within a subject (i.e. a higher intention level than his own average level) and differences in intention between subjects (i.e. subjects with higher intention levels *v.* subjects with lower intention levels) are linked to health factor. These findings align with the theory of planned behaviour and socio-ecological models, in which intention serves as a significant predictor of behaviour^(23,58). This is also supported by Inauen *et al.*, who were the first to examine the relationship between intention and snacking behaviour while disentangling the within- and between-person level⁽⁵⁹⁾. They found strong intention-behaviour associations at the within-person level, aligning with the assumptions of many health behaviour theories^(60,61). Besides, no significant intention-behaviour relations were found at the between-person level, supporting previous findings of an intention-behaviour gap^(19,62,63). The contradiction in literature may be attributed to the methodology and analysis employed because intention was mostly assessed only once at baseline.

In the current study, observing someone else eating was associated with a less healthy snack choice at both the within- and between-person level, indicating that seeing other people eat plays an important role in the snack behaviour in older adults. In other words, the association is not only present at a group level (i.e. across individuals in the study) but also at an individual level (i.e. within one participant). Other research suggests that observing someone else eating has a greater impact on the consumption of unhealthy snacks compared with healthy snacks⁽³²⁾. Both in previous studies and in the current study, the snacking behaviour of the opposing individual remains mostly unknown (e.g. the type of snack they ate), making it difficult to draw definitive conclusions on the mechanisms behind social modelling and the health factor of the snack. In contrast, the results of the current study also show that social modelling was not related to the decision to snack or to portion size. This finding contradicts other research suggesting that observing someone else eating is linked to a higher likelihood of snacking^(19,20,24). This is noteworthy, as social modelling is a well-established determinant, which is demonstrated in research involving numerous demographic groups⁽⁶⁴⁾. A systematic review on psychosocial factors influencing eating behaviours of older adults confirms that solitary eating is consistently negatively associated with the quantity of food consumed^(27,65). Furthermore, it is associated with an increased probability of having a lower BMI⁽⁶⁵⁻⁶⁷⁾, lower dietary variety^(67,68), lower consumption of fruits and vegetables^(66,68,69) and an elevated tendency to skip meals⁽⁶⁶⁾. It may be possible that the association between observing someone else eating and choosing to eat a snack as well as portion size may be weaker in older adults compared with other age groups, since older adults exhibit more habitual behaviours and eating patterns, making them less susceptible to social modelling⁽³²⁾. Further research should clarify the specific role of social modelling on older adults'



snacking behaviour. Some considerations can be taken into account: (1) Is the effect size of social modelling different when their eating partner or other surrounding people eat healthy or non-preferred foods compared with unhealthy or preferred foods?, (2) Are habitual snacking patterns responsible for conformity effects between eating companions? and (3) In a scenario in which both co-eaters are free to choose the type or amount of food to consume, which person is modelling and which person is being modelled?

When someone's feelings of cheerfulness were higher than their own average feelings of cheerfulness, they tended to make less healthy snack choices. Yet, no significant associations with health factor were found for negative emotions. These findings partially align with prior research reporting that older adults did not perceive emotions as important barriers or motivators in making food choices⁽²⁵⁾. This suggests that while emotional states may influence dietary behaviours, older adults might not be fully aware of this influence⁽²⁷⁾. Another study indicated that positive affect was associated with healthier diets, but that negative affect was not significantly related to dietary quality⁽²⁶⁾. Though this seems contradictory to the results of the current study, it is important to note that previous research is based at the between-person level. Contrarily, our findings also differ from previous research. Research on emotional eating in adults, characterised by overeating in response to negative emotions, indicates that negative emotions are associated with weight gain⁽⁷⁰⁾. On the contrary, a study within older adults demonstrated that negative emotions negatively affect dietary intake, with feelings of loneliness being linked to malnutrition⁽²⁴⁾. The foregoing suggests a nuanced relationship between eating behaviours and emotions across the lifespan, despite no associations being found at the between- or within-subject level with decision to snack or portion size in the current study. Discrepancies could arise from the interchangeable use of the terms 'mood' and 'emotion'⁽⁷¹⁾. Emotions and moods have distinct definitions; emotions refer to short-term affective responses to the appraisal of specific stimuli⁽⁷²⁻⁷⁴⁾, while moods are more prolonged psychological states that may persist independently of external triggers⁽⁷⁵⁾. Additionally, the discrepancies in findings may also be partially attributed to differences in methodologies employed, with only the present study using EMA to capture emotions at multiple time points. Loneliness, for example, is a longer lasting and more stable mood compared with cheerfulness that tends to be of more intense but short duration, and therefore is more suitable to be queried via EMA.

The results of the current study show that there is both a within-subject and between-subject association between the visibility of the snack and portion size. In other words, if a snack is visible to the person compared with when the snack is not visible for that same person (i.e. within-subject) or if a snack is visible compared with when it is not visible in general (i.e. between-subject), the portion size of the snack increases. These results are in line with previous research, although, only between-subject associations were taken into account in these previous studies^(19,20). It was observed that being exposed to food increased subjective experiences of hunger or desire to eat⁽⁷⁶⁾. This finding indicates that the presence of visible food can serve as a stimulus for eating a larger portion of the snack.

However these larger portion sizes should only be considered as negative when this is not in line with the nutritional requirements of the individual. With this in mind, future interventions should address visibility in an appropriate way for a specific individual (e.g. eliminating unhealthy snacks from the view of a person who is overweight).

The lack of significant results for the other determinants also deserves attention. There are different possible explanations for the lack of significant results and a distinction must be made between the within-person variance and the between-person variance. On the one hand, the lack of within-person associations detected might be attributed to the limited intra-individual variability of some of the reported determinants within days. Online Supplementary file 3 shows in detail how a determinant fluctuates for each participant individually. The majority of the participants reported limited variation within days for the negative emotions (i.e. loneliness, insecurity, anxiousness, irritation, feeling down). For example, on 62% (i.e. 211 out of 339) of the total amount of triggers studied, the determinant loneliness did not fluctuate within a participant. Consequently, it is possible that within-person variability in these determinants is not relevant for the subgroup of older adults. On the other hand, the fluctuation of determinants is also rather limited at the between-person level. In the current study, the standard deviation of negative emotions, loneliness, insecurity, anxiousness, irritation and feeling down was overall low (i.e. SD ranged between 0.6 (negative emotions) to 0.9 (irritation)). A slightly higher variation was found for positive emotions, cheerfulness, relaxation, enthusiasm and satisfaction (i.e. SD ranged between 0.8 (positive emotions) to 1.1 (relaxation)), but remained rather limited. Not only the variation of these determinants is quite limited, but also the reported scores were overall low (i.e. the mean score ranged between 1.4 (anxiousness) and 1.6 (irritation) on 7). Consequently, the level of these determinants might have been too low to influence participant's snacking behaviour. Our findings could be explained in light of the paradox of emotional well-being in ageing⁽²⁸⁾. With ageing, physical health, strength, cognitive agility and social networks decline. Yet, on average, emotional well-being is somehow maintained or even improved as older adults age⁽²⁹⁾. This is because older adults excel in regulating their emotions compared with younger adults, enabling them to more effectively diffuse negative feelings and situations^(30,31). In conclusion, there may potentially be no relationship between the fluctuation in someone's emotions and snacking behaviour. However, further investigation in a more representative target group is warranted to clarify the findings from the current study and better understand the role of emotions in snack consumption among older adults.

Neither significant within- nor between-subject associations were found for the determinants and the outcome variable decision to snack. This lack of association may be attributed to the tendency of older adults to not base their decision to snack on factors such as emotions, social modelling or the visibility of snacks. Older adults may exhibit more habitual snacking behaviours, making them less susceptible to the influence of determinants⁽⁵⁵⁾. In addition, eating a snack is not necessarily considered negative, as long as one chooses a healthy snack with an adequate portion size. The question arises if the outcome

variable decision to snack may not be suited to interventions that offer support at specific moments (e.g. JITAD).

The present study has several strengths. A first strength is the use of the innovative monitoring strategy EMA. It involves repeated sampling of experiences during people's everyday lives, which maximises ecological validity and minimises recall bias⁽⁴¹⁾. It also allowed us to identify time-dependent as well as context-dependent variations of determinants. More detailed results on the compliance rate in this study are reported elsewhere (Compernelle *et al.*, under review). A second strength is that this study fills an important gap in literature, since studies on snacking behaviour and the determinants are very limited, especially in older adults. This study encompasses a comprehensive assessment of various food items, with both unhealthy and healthy snacks, and their portion sizes. Future studies, however, could delve deeper into the needs of specific groups, such as malnourished, overweight or obese older adults. Lastly, according to our knowledge, this is the first study examining both the within- and between-person associations between snacking behaviour and determinants. The focus on the within-individual processes demonstrates the uniqueness of the paper. This is an important perspective, as not everyone follows the same behaviour (change) processes⁽⁷⁷⁾.

However, the study also holds limitations. First, the study results may have limited generalisability due to nonprobability sampling, as this sample may not accurately represent the population of older adults (e.g. a high number of highly educated participants) and may have left out more socially isolated individuals. It is therefore recommended that future studies adopt a more random sampling approach to obtain a more heterogeneous study sample to be able to generalise the study results to a wider population of older adults. Second, a hybrid design was adopted (i.e. utilising both pencil-paper diaries and smartphone-based questions for data collection). However, it is crucial to acknowledge an important disadvantage associated with the use of paper diaries⁽⁵⁷⁾. Ensuring accurate combination of time- and event-based datasets relies heavily on precise time stamps recorded in the paper diaries. Another concern is the potential occurrence of backfilling, when participants fail to complete the questionnaire at the required times and hastily fill them in before returning to the researchers^(78,79). This practice could have detrimental consequences for the validity of the data. In light of these limitations, researchers could offer participants the flexibility to choose between paper-and-pencil or smartphone-based methods for event-based diary entries in future research. Moreover, when utilising a smartphone-based approach, there may be merit in developing clear and tailored manuals and training sessions in collaboration with the target population. This may enhance participants' digital literacy, thereby optimising data collection efficiency and reliability. Third, using a consistent definition of 'snack' provides both an advantage and a disadvantage. While it offers a clear delineation of the concept, it excludes food immediately following the main meal and all beverages. For solid foods, valuable information linked to main meals, such as post-meal dessert consumption has been not taken into account. For beverages, the consumption patterns differ from those of solid foods due to unique consumer psychology. For instance,

alcoholic drinks enjoyed socially or alone, coffee consumed either away from home or at home are typically not consumed in response to hunger and are not commonly considered snacks⁽¹²⁾. In the current study, researchers deemed it important to minimise participant burden and reduce confusion surrounding snack definitions, especially within this vulnerable group where the use of EMA via smartphones was already challenging⁽¹³⁾. Putting that aside, future research could indeed focus on beverage consumption among older adults as there is still a gap in the literature on this topic. Fourth, in other EMA research, participants reported moderate reactivity in becoming more aware of their eating behaviour during the study (i.e. mean = 54.94 on a scale from 0–100), and low to moderate reactivity regarding the influence of morning assessments on the eating behaviour on the respective days (mean = 27.22 on a scale from 0–100)⁽⁸⁰⁾. The question about intention for healthy snacking could have induced the craving for snacks, although we cannot say this with certainty since reactivity was not queried in the current study. Fifth, due to the matching of the time-based and event-based data sets, a number of measurements have been lost, possibly resulting in insufficient power for the outcome variables health factor and portion size. Future research should consider a larger sample size or longer measurement period rather than increasing the number of triggers per day, as the number of snacking moments in a day is limited. Lastly, possible confounding factors, such as preceding eating moments, other emotions, or other unmeasured variables, were not accounted for in this study. Hence, it would be interesting to further examine possible confounding factors in future EMA research.

The insights gained from understanding the determinants can inform future health interventions and can specifically inform decisions regarding which determinants to focus on to promote health behaviours. Based on our findings, it can be suggested that these interventions should focus on the determinants intention and social modelling. Moreover, in the future, more personalised interventions (e.g. JITAD) could be designed to provide appropriate support to participants at the right moments to further enhance their effectiveness in changing behaviour. JITAD could for instance focus on determinants like cheerfulness because these vary the most intra-individually (within-person). Besides, the results for the other determinants (i.e. loneliness, relaxation, enthusiasm, satisfaction, insecurity, anxiousness, irritation and feeling down), showing limited variation over days and within days and a lack of significant associations with snacking behaviour, imply that interventions targeting these determinants might have limited impact. All in all, as this study was primarily exploratory in nature, more research, which addresses a more representative sample size, is required to formulate clearer guidelines for future health interventions.

Conclusions

The present study revealed that levels of intention, social modelling and cheerfulness within individuals were associated with health factor. Having a higher intention level than someone's average intention level leads to healthier snacking while seeing someone in the environment compared with not



seeing someone in the environment and a higher cheerfulness level than someone's average cheerfulness level promotes unhealthier choices. Additionally, intention and social modelling show a between-person association with health factor, where higher intention aligns with healthier choices, and higher levels of social modelling lead to unhealthier choices. Furthermore, visibility affects portion size both on a within- and between-person level. The portion size increases when the snack is visible. Besides, no associations were found between the determinants and the decision to snack. Through the implementation of EMA, this study has generated novel insights into the within-person associations between time-varying determinants and snacking, thereby offering valuable implications for future research. Such information holds the potential for incorporation into JITAI, allowing for personalized tailoring, more effective promotion of healthier snacking behaviours and thus pursuing the challenge of healthy ageing.

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The authors declare none.

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Supplementary material

For supplementary material accompanying this paper visit <https://doi.org/10.1017/S0007114524001004>

References

1. The World Bank (2022 Revision) United Nations Population Division: World Population Prospects. <https://data.worldbank.org/indicator/SP.POP.65UP.TO.ZS> (accessed 14 April 2023).
2. He W, Goodkind D & Kowal P (2016) *An Aging World: 2015* (UC Bureau, editor). Washington, DC: U.S. Government Publishing Office.
3. WHO (2020) Healthy Ageing and Functional Ability. <https://www.who.int/news-room/questions-and-answers/item/healthy-ageing-and-functional-ability> (accessed 23 November 2023).
4. Black M & Bowman M (2020) Nutrition and healthy aging. *Clin Geriatr Med* **36**, 655–669.
5. Volkert D (2013) Malnutrition in older adults – urgent need for action: a plea for improving the nutritional situation of older adults. *Gerontology* **59**, 328–333.
6. Zizza CA, Tayie FA & Lino M (2007) Benefits of snacking in older Americans. *J Am Dietetic Assoc* **107**, 800–806.
7. Mena B, Ashman H, Dunshea FR, *et al.* (2020) Exploring meal and snacking behaviour of older adults in Australia and China. *Foods (Basel, Switzerland)* **9**, 426.
8. Skoczek-Rubinska A & Bajerska J (2021) The consumption of energy dense snacks and some contextual factors of snacking may contribute to higher energy intake and body weight in adults. *Nutr Res* **96**, 20–36.
9. Marzetti E, Calvani R, Tosato M, *et al.* (2017) Sarcopenia: an overview. *Aging Clin Exp Res* **29**, 11–17.
10. Nazri NSM, Vanoh D & Soo KL (2022) Natural food for Sarcopenia: a narrative review. *Malays J Med Sci* **29**, 28–42.
11. de Graaf C (2000) *Nutritional Definitions of the Meal, Dimensions of the Meal: The Science, Culture, Business and Art of Eating*. Gaithersburg, MD: Aspen Publishers, Inc.
12. Johnson GH & Anderson GH (2010) Snacking definitions: impact on interpretation of the literature and dietary recommendations. *Crit Rev Food Sci Nutr* **50**, 848–871.
13. Yao L, Yang Y, Wang Z, *et al.* (2023) Compliance with ecological momentary assessment programmes in the elderly: a systematic review and meta-analysis. *BMJ Open* **13**, e069523.
14. Schlettwein-Gsell D, Decarli B & De Groot L (1999) Meal patterns in the SENECA study of nutrition and the elderly in Europe: assessment method and preliminary results on the role of the midday meal. *Appetite* **32**, 15–22.
15. Vatanparast H, Islam N, Masoodi H, *et al.* (2020) Time, location and frequency of snack consumption in different age groups of Canadians. *Nutr J* **19**, 85.
16. Andersson J, Nydahl M, Gustafsson K, *et al.* (2003) Meals and snacks among elderly self-managing and disabled women. *Appetite* **41**, 149–160.
17. Higgins MM (2007) Food and nutrition professionals can help older adults improve dietary practices. *J Am Dietetic Assoc* **107**, 806–807.
18. Meltzer HM, Brantsaeter AL, Trolle E, *et al.* (2019) Environmental sustainability perspectives of the Nordic diet. *Nutrients* **11**, 2248.
19. Elliston KG, Ferguson SG & Schuz B (2017) Personal and situational predictors of everyday snacking: an application of temporal self-regulation theory. *Br J Health Psychol* **22**, 854–871.
20. Elliston KG, Ferguson SG, Schuz N, *et al.* (2017) Situational cues and momentary food environment predict everyday eating behavior in adults with overweight and obesity. *Health Psychol* **36**, 337–345.
21. Reichenberger J, Kuppens P, Liedlgruber M, *et al.* (2018) No haste, more taste: an EMA study of the effects of stress, negative and positive emotions on eating behavior. *Biol Psychol* **131**, 54–62.
22. Eskelinen K, Hartikainen S & Nykanen I (2016) Is loneliness associated with malnutrition in older people? *Int J Gerontol* **10**, 43–45.
23. Glanz K, Rimer B & Viswanath K (2015) *Health Behaviour: Theory, Research and Practice*, 5th ed. San Francisco: Jossey-Bass.
24. Hess JM, Jonnalagadda SS & Slavin JL (2016) What is a snack, why do we snack, and how can we choose better snacks? A review of the definitions of snacking, motivations to snack, contributions to dietary intake, and recommendations for improvement. *Adv Nutr* **7**, 466–475.

25. Locher JL, Ritchie CS, Roth DL, *et al.* (2009) Food choice among homebound older adults: motivations and perceived barriers. *J Nutr Health Aging* **13**, 659–664.
26. Whitehead BR (2017) Health behaviors in older adults: considering age, affect, and attitudes. *J Health Psychol* **22**, 1652–1657.
27. Walker-Clarke A, Walasek L & Meyer C (2022) Psychosocial factors influencing the eating behaviours of older adults: a systematic review. *Ageing Res Rev* **77**, 101597.
28. Mather M (2012) The emotion paradox in the aging brain. *Ann Ny Acad Sci* **1251**, 33–49.
29. Charles ST (2010) Strength and vulnerability integration: a model of emotional well-being across adulthood. *Psychol Bull* **136**, 1068–1091.
30. Carstensen LL, Pasupathi M, Mayr U, *et al.* (2000) Emotional experience in everyday life across the adult life span. *J Pers Soc Psychol* **79**, 644–655.
31. Hay EL & Diehl M (2011) Emotion complexity and emotion regulation across adulthood. *Eur J Ageing* **8**, 157–168.
32. Cruwys T, Bevelander KE & Hermans RC (2015) Social modeling of eating: a review of when and why social influence affects food intake and choice. *Appetite* **86**, 3–18.
33. Cleobury L & Tapper K (2014) Reasons for eating ‘unhealthy’ snacks in overweight and obese males and females. *J Hum Nutr Dietetics: Offic J Br Dietetic Assoc* **27**, 333–341.
34. Phan UTX & Chambers E (2018) Motivations for meal and snack times: three approaches reveal similar constructs. *Food Qual Prefer* **68**, 267–275.
35. Verhoeven AAC, Adriaanse MA, de Vet E, *et al.* (2015) It’s my party and I eat if I want to. Reasons for unhealthy snacking. *Appetite* **84**, 20–27.
36. de Gooijer FJ, van Kraaij A, Fabius J, *et al.* (2023) Assessing snacking and drinking behavior in Real-Life Settings: validation of the SnackBox technology. *Food Qual Prefer* **112**, 105002.
37. Berkman ET, Giuliani NR & Pruitt AK (2014) Comparison of text messaging and paper-and-pencil for ecological momentary assessment of food craving and intake. *Appetite* **81**, 131–137.
38. Stone AA, Shiffman S, Schwartz JE, *et al.* (2002) Patient non-compliance with paper diaries. *BMJ (Clinical Res Ed)* **324**, 1193–1194.
39. Maes I, Mertens L, Poppe L, *et al.* (2022) The variability of emotions, physical complaints, intention, and self-efficacy: an ecological momentary assessment study in older adults. *PeerJ* **10**, e13234.
40. Spanakis G, Weiss G, Boh B, *et al.* (2016) Network analysis of ecological momentary assessment data for monitoring and understanding eating behavior. *Lect Notes Comput Sci* **9545**, 43–54.
41. Mauger A & Barchitta M (2019) A systematic review of ecological momentary assessment of diet: implications and perspectives for nutritional epidemiology. *Nutrients* **11**, 2696.
42. Nahum-Shani I, Hekler EB & Spruijt-Metz D (2015) Building health behavior models to guide the development of just-in-time adaptive interventions: a pragmatic framework. *Health Psychol* **34**, 1209–1219.
43. Peersman T (2014) *Portiegrootboek*, vol. 3. Waasmunster, Belgium: Valetudo Consulting.
44. UK Department of Health (2011) *Nutrient Profiling Technical Guidance*. London: UK Department of Health.
45. Superior Health Council Belgium (2005) *Sizes and Weights*, 2nd ed. Belgium: Superior Health Council Belgium.
46. Degroote L, Van Dyck D, De Bourdeaudhuij I, *et al.* (2020) Acceptability and feasibility of the mHealth intervention ‘MyDayPlan’ to increase physical activity in a general adult population. *BMC Public Health* **20**, 1032.
47. Liao Y, Skelton K, Dunton G, *et al.* (2016) A systematic review of methods and procedures used in ecological momentary assessments of diet and physical activity research in youth: an adapted STROBE Checklist for Reporting EMA Studies (CREMAS). *J Med Internet Res* **18**, e151.
48. O’Brien ST, Dozo N, Hinton JDX, *et al.* (2023) SEMA3: A Free Smartphone Platform for Daily Life Surveys. *PsxArXiv*. <https://doi.org/10.31234/osf.io/cnar8> (accessed August 2021).
49. Dunton GF (2017) Ecological momentary assessment in physical activity research. *Exerc Sport Sci Rev* **45**, 48–54.
50. Delespaul P (1995) *Assessing Schizophrenia in Daily Life. The Experience Sampling Method*. Maastricht: Maastricht University Press.
51. R-Project (2023) R: A Language and Environment for Statistical Computing. <https://www.r-project.org/> (accessed July 2023).
52. Bates D, Maechler M, Bolker B, *et al.* (2023) lme4: Linear Mixed-Effects Models using ‘Eigen’ and S4. <https://CRAN.R-project.org/package=lme4>
53. Babyak M (2004) What you see may not be what you get: a brief, nontechnical introduction to overfitting in regression-type models. *Psychosom Med* **66**, 411–421.
54. Peduzzi P, Concato J, Kemper E, *et al.* (1996) A simulation study of the number of events per variable in logistic regression analysis. *J Clin Epidemiol* **49**, 1373–1379.
55. Vatanparast H, Islam N, Patil RP, *et al.* (2019) Snack consumption patterns among Canadians. *Nutrients* **11**, 1152.
56. McNeish DM & Stapleton LM (2016) The effect of small sample size on two-level model estimates: a review and illustration. *Educ Psychol Rev* **28**, 295–314.
57. Myin-Germeys I & Kuppens P (2022) *The Open Handbook of Experience Sampling Methodology: A Step-by-Step Guide to Designing, Conducting, and Analyzing ESM Studies*, 2nd ed. Leuven: Center for Research on Experience Sampling and Ambulatory Methods Leuven.
58. Mcleroy KR, Bibeau D, Steckler A, *et al.* (1988) An ecological perspective on health promotion programs. *Health Educ Quart* **15**, 351–377.
59. Inauen J, Shrout PE, Bolger N, *et al.* (2016) Mind the Gap? An intensive longitudinal study of between-person and within-person intention-behavior relations. *Ann Behav Med* **50**, 516–522.
60. Ajzen I (1991) The theory of planned behavior. *Organ Behav Hum Dec* **50**, 179–211.
61. Schwarzer R (2008) Modeling health behavior change: how to predict and modify the adoption and maintenance of health behaviors. *Appl Psychol-Int Rev* **57**, 1–29.
62. Rhodes RE & Dickau L (2012) Experimental evidence for the intention-behavior relationship in the physical activity domain: a meta-analysis. *Health Psychol* **31**, 724–727.
63. Webb TL & Sheeran P (2006) Does changing behavioral intentions engender behaviour change? A meta-analysis of the experimental evidence. *Psychol Bull* **132**, 249–268.
64. Herman CP, Koenig-Nobert S, Peterson JB, *et al.* (2005) Matching effects on eating: do individual differences make a difference? *Appetite* **45**, 108–109.
65. Wham CA & Bowden JA (2011) Eating for health: perspectives of older men who live alone. *Nutr Diet* **68**, 221–226.
66. Tani Y, Kondo N, Takagi D, *et al.* (2015) Combined effects of eating alone and living alone on unhealthy dietary behaviors, obesity and underweight in older Japanese adults: results of the JAGES. *Appetite* **95**, 1–8.
67. Kimura Y, Wada T, Okumiya K, *et al.* (2012) Eating alone among community-dwelling Japanese elderly: association with depression and food diversity. *J Nutr Health Aging* **16**, 728–731.





68. Schnettler B, Lobos G, Lapo MD, *et al.* (2017) Satisfaction with life and food-related life in Ecuadorian older adults-related. *Nutr Hosp* **34**, 65–72.
69. Rugel EJ & Carpiano RM (2015) Gender differences in the roles for social support in ensuring adequate fruit and vegetable consumption among older adult Canadians. *Appetite* **92**, 102–109.
70. Barak RE, Shuval K, Li Q, *et al.* (2021) Emotional eating in adults: the role of sociodemographics, lifestyle behaviors, and self-regulation-findings from a US National Study. *Int J Env Res Pub Health* **18**, 1744.
71. Köster EP & Mojet J (2015) From mood to food and from food to mood: a psychological perspective on the measurement of food-related emotions in consumer research. *Food Res Int* **76**, 180–191.
72. Frijda NH (1999) *Emotions and Hedonic Experience, Well-Being: The Foundations of Hedonic Psychology*. New York: Russell Sage Foundation.
73. Matthews G, Deary IJ & Whiteman MC (2003) *Personality Traits*, 2nd ed. Cambridge: Cambridge University Press.
74. Rolls ET (1999) *The Brain and Emotion*. Oxford: Oxford University Press.
75. Thayer RE (1990) *The Biopsychology of Mood and Arousal*. Oxford: Oxford University Press.
76. Ferriday D & Brunstrom JM (2011) 'I just can't help myself': effects of food-cue exposure in overweight and lean individuals. *Int J Obes (Lond)* **35**, 142–149.
77. Heino MTJ, Knittle K, Noone C, *et al.* (2021) Studying behaviour change mechanisms under complexity. *Behav Sci (Basel)* **11**, 77.
78. Smyth JM & Stone AA (2003) Ecological momentary assessment research in behavioral medicine. *J Happiness Stud* **4**, 35–52.
79. Stone AA, Broderick JE, Schwartz JE, *et al.* (2003) Intensive momentary reporting of pain with an electronic diary: reactivity, compliance, and patient satisfaction. *Pain* **104**, 343–351.
80. Pannicke B, Kaiser T, Reichenberger J, *et al.* (2021) Networks of stress, affect and eating behaviour: anticipated stress coping predicts goal-congruent eating in young adults. *Int J Behav Nutr Phys Act* **18**, 9.