Relationships between food groups and eating time slots according to diabetes status in adults from the UK National Diet and Nutrition Survey (2008–2017)

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2 ABSTRACT

Time of eating has been shown to be associated with diabetes and obesity but little is known 3 about less healthy foods and specific time of their intake over the 24 hours of the day. In this study we aimed to identify potential relationships between foods and their eating time, and see whether these associations may vary by diabetes status. The National Diet and Nutrition Survey (NDNS) including 6802 adults (age > 19 years old) collected 749,026 food recordings 7 by a 4-day-diary. The contingency table cross-classifying 60 food groups with 7 pre-defined eating time slots (6-9am, 9am-12pm, 12-2pm, 2-5pm, 8-10pm, 10pm-6am) was analyzed by Correspondence Analysis (CA). CA biplots displaying the associations were generated for all 10 adults and separately by diabetes status (self-reported, pre-diabetes, undiagnosed-diabetes, and 11 12 non-diabetics) to visually explore the associations between food groups and time of eating across diabetes strata. For selected food groups, odds ratios (OR, 99% confidence intervals, CI) were derived of consuming unhealthy foods at evening/night (8pm-6am) vs. earlier time in the day, by logistic regression models with generalized estimating equations. The biplots suggested positive 15 associations between evening/night and consumption of puddings, regular soft drinks, sugar 16 confectioneries, chocolates, beers, ice cream, biscuits, and crisps for all adults in the UK. The OR 17 (99% Cls) of consuming these foods at evening/night were respectively 1.43 (1.06, 1.94), 1.72 (1.44, 2.05), 1.84, (1.31, 2.59), 3.08, (2.62, 3.62), 7.26, (5.91, 8.92), 2.45, (1.84, 3.25), 1.90, (1.68, 1.90)19 2.16), 1.49 (1.22, 1.82) vs. earlier time in the day adjusted for age, sex, body mass index, and 20 social-economic levels. Stratified biplots found that sweetened beverages, sugar-confectioneries appeared more strongly associated with evening/night among un-diagnosed diabetics. Foods consumed in the evening/night time tend to be highly processed, easily accessible, and rich in added sugar or saturated fat. Individuals with undiagnosed diabetes are more likely to consume 24 unhealthy foods at night. Further longitudinal studies are required to ascertain the causal direction of the association between late-eating and diabetes status.

27 Keywords: Chrononutrition, time of eating, correspondence analysis, the UK National Diet and Nutrition Survey, nutrition epidemiology,

28 diabetes

INTRODUCTION

The timing of energy intake has been shown to be associated with obesity and diabetes. (Almoosawi et al., 2016) Specifically, eating late at night or having a late dinner was found to be related to higher 30 risk of obesity (Xiao et al., 2019; Yoshida et al., 2018), hyperglycemia (Nakajima and Suwa, 2015), 31 metabolic syndrome (Kutsuma et al., 2014), diabetes (Mattson et al., 2014), and poorer glycemic control 32 among diabetics (Sakai et al., 2017). However, the relationship between food choice and the time of 33 food consumption during the day is left largely unknown. Shiftworkers have an increased risk of obesity 34 35 (Balieiro et al., 2014; Barbadoro et al., 2013), and diabetes (Pan et al., 2011), possibly due to limited availability of healthy food choice during their night shifts (Bonnell et al., 2017; Balieiro et al., 2014). 36 Previous survey data from the UK National Diet and Nutrition Survey Rolling Programme (NDNS RP) 37 found that overall, 3.4% of men and 2.3% of women aged 19-64 had fasting glucose concentrations above 38 the clinical cut-off for diabetes (≥ 7 mmol/L). Moreover, the proportion of men with undiagnosed diabetes 39 increased with age to over 20% in the UK population (Almoosawi et al., 2014). Identifying those unhealthy 40 foods that might be chosen during late night time would be helpful when guiding people to change their 41 eating habit for the purpose of either weight loss or glycemic control. Dietary diary recordings from NDNS 42 RP surveys can provide detailed food choice data for exploration of the relationships between food groups 43 and their time of consumption in the general population. 44

In this study, we aimed to describe the relationship between food groups and the time of day when they were consumed, and how such relationships may vary by status of type 2 diabetes using the data published by the NDNS RP from 2008 to 2017 as this survey includes diet diaries providing detailed information on the time of day of food intake.

METHODS

6802 adults (2810 men and 3992 women) and 749026 food recordings collected by the NDNS RP 2008-17 were analyzed in the current study (MRC Elsie Widdowson Laboratory and NatCen Social Research, 2018). The survey comprised a cross-section representative sample of the UK adult population taken over the 51 52 period 2008-2017. The sample was randomly drawn from a list of all addresses in the UK, clustered into postcode sectors. Details of the rationale, design and methods of the survey can be found in the previously 53 published official study reports (Bates et al., 2014; Roberts et al., 2018). A four-day food diary method was 54 used in the NDNS RP to collect the detailed food items and their time of consumption from participants. 55 Comparison between the food diary method and a repeated 24-hour recall questionnaire was performed in 56 a subset of study sample prior to the launch of the NDNS RP in 2008 and found that they were similar 57 in terms of response rate as well as the ability to collect correct nutrition intake data. And the four-day 58 food diary methos was adopted because it is considered to be more flexible and adaptable to cover wide 59 population age range in the survey. More details can be found in the Appendix A of the official NDNS RP study report (Bates et al., 2014; Roberts et al., 2018). Furthermore, the same food diary methods is 61 actually used in large studies conducted in the UK, such as the the MRC National Survey of Health and 62 Development (NSHD) (1946 British Birth Cohort) (Price et al., 1997), the EPIC Norfolk Study (Bingham et al., 2001), the UK Women's Cohort Study in Leeds (Cade et al., 2004), and the Avon Longitudinal Study of Parents and Children (ALSPAC) cohort (Glynn et al., 2005).

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In the food diary recordings, time of the day was categorized into 7 slots: 6-9 am, 9-12 noon, 12-2 pm, 2-5 66 67 pm, 5-8 pm, and 10 pm - 6 am. Foods recorded were classified into 60 standard food groups with 1 to 10 subgroups each: the details are given in Appendix R of the NDNS official report (NatCen Social Research, 68 69 MRC Elsie Widdowson Laboratory, University College London. Medical School., 2018). We focused on 70 the 60 standard food groups in the current analysis. Diabetes status was defined as: 1) healthy if fasting glucose was lower than 6.10 (mmol/L), hemoglobin A1c (HbA1c) was less than 6.5 (%), and without 71 72 self-reported diabetes and treatment for diabetes (n = 2626); 2) pre-diabetic if fasting glucose was lower 73 between 6.10 and 6.99 (mmol/L, inclusive) but without self-reported diabetes and without treatment for 74 diabetes (n = 133); 3) undiagnosed diabetic if either fasting glucose was higher or equal to 7.00 (mmol/L) 75 or HbA1c higher or equal to 6.5 (%) but without self-reported diabetes and treatment for diabetes (n = 99); 4) diabetic if participant had self-reported diabetes or was under treatment for diabetes (n = 227). 76 77 Consequently, there was also a large number of adults (3717 adults of whom 1519 men and 2198 women) whose diabetes status did not fall in one of above categories and could not thus be confirmed; these were retained in the whole sample (unstratified) analyses. In addition, the National Statistics Socio-economic 79 Classification (Rose and Pevalin, 2005) was applied in the survey and accordingly, the socio-economic 80 status of participants was classified in one of 8 categories.

Correspondence analysis (CA) (Greenacre, 2017; Chapman et al., 2017; Palla et al., 2020) was used as a tool for data mining, visualization and hypotheses generation using half of the randomly selected NDNS diary entries data. Specifically, the contingency table generated by cross-tabulating 60 food groups and 7 time slots were analyzed by CA. Through CA, the 60 categories of standard foods and the 7 time slots were projected on biplots, i.e. onto two dimensional plots that could jointly contain large percentage of the χ^2 deviation (or inertia) of the contingency table. Biplots that graphically show the association between time of day and food groups were derived for all adults and separately according to their diabetes status. To account for the hierarchical structure of the data (food recorded by the same individuals who lived within the same area/sampling units) and to calculate population average odds ratios (OR), logistic regression models with generalized estimating equations (GEE) were subsequently used to test the associations that were first suggested by visual inspection of biplots generated by CA, using the remaining half of the diary entries data. The marginal ORs and their 99% confidence intervals (CI) were derived of consuming unhealthy food groups (selected by CA) later in the day (8 pm - 6 am, i.e. in the evening and night) compared to earlier in the day (in the morning or afternoon). CA and biplots were conducted and generated by the following packages under R environment (R Core Team, 2019): FactoMineR, factoextra, ggplot2, ggrepel (Lê et al., 2008; Kassambara and Mundt, 2019; Wickham, 2016; Slowikowski, 2019) Logistic regression models with GEE were performed with SAS procedure GENMOD (SAS Institute, 2013) adjusted for age, sex, and socio-economic levels, which were deemed the main potential confounders of the associations.

RESULTS

The dataset consisted of 2810 (41.3%) men and 3992 (58.7%) women aged older than or equal to 19 years old with the mean age of 49.9 years (standard deviation, SD = 17.6). Of these individuals 22.6 % were current smokers, 24.3 % were past smokers. The average body mass index (BMI) was 27.7 kg/m² (SD = 5.41). Among the food recordings collected (n = 749026), 56.9% were recorded during traditional breakfast (6 am - 9 am: 14.3%), lunch (12 noon - 2 pm: 18.5%), or dinner (5 pm - 8 pm: 24.1%) time slots. Table 1. shows the top 37 food groups that contributed to 90% of the total calories consumed by adults in NDNS RP. These food groups accounted for 478028 of the total diary entries (63.8 %). The random process split

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the whole set of food recordings into a hypothesis generating dataset of 374682 and a testing dataset of 374344 entries. 109

110 Figure 1-5 present the CA biplots that visually summarize the associations between 60 food groups and the time of their consumption in the entire sample and then stratifying by their diabetes status. In Figure 1, 111 the horizontal axis explains 68.9 % of the association structure (inertia) between food and time while the 112 vertical axis reflects 15.3 % of the same relationship. Therefore, a total of 84.2 % of the inertia between food and time were captured in this figure which shows a visual summary of how those two categorical 114 variables are related. Specifically, time slots later than 8 pm are shown in the upper side of the plot closer to 115 alcoholic products (beers etc.) or highly processed/energy-dense foods (sugar confectioneries, chocolates, biscuits, regular softdrinks, ice cream, crisps); times earlier than noon appear in the left hand side together 117 with typical breakfast foods (cereals, milk, bread, etc.). 118

To visualize the potentially different associational patterns between food groups choice and time slots 120 according to diabetes status, Figure 2-5 display the CA biplots in subsets of the data. Depending on different diabetes status, these biplots explained between 76.3% and 84.1% of the inertia in the data. Similarly to the biplot created from the total sample (Figure 1), later time in the day (8 pm and later) are shown in the upper side of each figure and suggested an association with the alcoholic beverages and highly processed or energy-dense food groups. Additionally, some food groups and time slots also flagged up associations potentially different by diabetes status. For example, puddings seemed to be closer to later time in the day among undiagnosed diabetics (Figure 4) while for diagnosed diabetic patients (Figure 3) they were closer to traditional dinner time (5 pm to 8 pm) or earlier in the day. Furthermore, sugar confectioneries/chocolates/biscuits/regular soft drinks appeared to be associated with later time in the day (8 pm or later) more strongly among undiagnosed diabetics (Figure 4) than the other participants.

130 Based on the findings suggested from Figure 1-5, we decided to focus on puddings, regular soft drinks, 131 confectioneries, chocolates, beers, ice cream, biscuits, crisps as these foods either showed a particularly strong association with time of the day or a different pattern of association across different strata of the 132 survey sample; hence, we tested the following null hypotheses using logistic regression models (adjusted 133 134 for age, sex, and socio-economic levels) with GEE: that the odds of consuming each selected food at later time of the day (8 pm - 6 am) is the same compared to earlier in the day; and the associations of the 135 above-mentioned food groups and time slots are the same among participants with different diabetes status 137 (i.e. no interaction between the time of food intake and diabetes status). The results are summarized in Table 2. 138

The listed food groups were found to have higher odds to be consumed between 8 pm and 6 am with higher 139 odds compared to earlier time. The OR (99% CIs) main effects of consuming these foods at evening/night 140 were for puddings 1.43 (1.06, 1.94), for regular soft drinks 1.72 (1.44, 2.05), for sugar confectioneries 1.84 141 142 (1.38, 2.69), for chocolates 3.08 (2.62, 3.62), for beers 7.26 (5.91, 8.92), for ice cream 2.45 (1.84, 3.25), for biscuits 1.90 (1.68, 2.16), for crisps 1.49 (1.22, 1.82) vs. earlier time. Opposite directions of the association 143 for puddings were detected across diabetes status: the ORs (99% CIs) of consuming puddings at night 144 145 time (8 pm or later) compared to earlier time were 1.55 (1.13, 2.15), 0.95 (0.17, 5.20), 1.82 (0.41, 8.03), and 0.63 (0.15, 2.66) for healthy, prediabetic, undiagnosed diabetic, and diabetic participants, respectively. 146 Furthermore, undiagnosed diabetic patients were found to have particularly high odds of consuming regular 147 soft drinks (OR: 2.82; 99% CI: 1.24, 6.43), and sugar confectioneries (OR: 10.61; 99% CI: 2.35, 47.04) 148 during night time periods compared to participants with other diabetes status. 149

DISCUSSION

The present study described the potential relationships between food groups and time of their consumption in a representative sample from the NDNS RP. Many unhealthy foods emerged from CA were found to be more likely to be consumed after 8 pm. These included alcoholic/sweetened beverages, chocolates and other foods rich in added sugars and saturated fats such as biscuits and ice cream. Foods chosen in the evening/night time slots tend to be highly processed and easily accessible. Specifically, undiagnosed patients might be at a higher risk of worsening their condition as they were found to have higher odds to choose a number of less healthy foods after 8 pm (sugar confectioneries, regular soft drinks) than diabetics and non-diabetics. Those foods might need to be targeted when designing intervention to those who might be at risk of being diabetics.

These findings are concerning considering previous research that have indicated that quality of macronutrient intake in the evening is likely to influence fasting glucose levels and glycaemic response to subsequent meals in the morning. (Wolever et al., 1988) More recently, a randomized controlled trial indicated that consuming carbohydrates at dinner irrespective of glycaemic index raised postprandial glucose response to breakfast producing what is known as a second meal effect (Haldar et al., 2020). Similar observation have been made by Nitta and colleagues who observed that eating sweet snacks post-dinner worsened glycaemic excursions in the evening and at subsequent breakfast (Nitta et al., 2019). Added to this is evidence that suggests that the late-night dinners induce post-prandial hyperglycemia in patients with type 2 diabetes and that interventions at this eating occassions can result in a profound impact on post-prandial glycaemia. On the balance of this evidence, targeting and improving the timing and quality of foods in evening eating occasions provides a unique opportunity to design intervention to those who might be at risk of being diabetics.

A compelling finding of our study is the observation that diabetes patients were found to be potentially controlling their choice of food groups such as avoiding puddings at night. However, higher odds of consuming alcoholic beverages and energy condensed foods such as chocolates and sugar confectioneries at night among individuals with diabetes suggests that their food choice might need further modifications.

Assessing the relationships between food groups and timing of eating by diabetes status can be considered as a first step towards identifying specific public health targets for behavior change/intervention. This is important as most current public health strategies and dietary recommendations do not provide targeted advice that takes into considerations specific eating occasions while targeted advice is more likely to result in sustainable behavioural change. Our findings are consistent with previous evidence that has found that both sweetened and alcoholic beverages are responsible for large portion of energy consumption at night in other populations (Hassen et al., 2018).

However, an important limitation in this study is the cross-sectional study design. The inability to assess the temporal relationship between timing of food intake and diabetes status means that a cause-effect relationship between time of unhealthy food intake and diabetes status cannot be established. Hence, further prospective studies are warranted to investigate the causal relationship between diabetes and both quality and timing of eating. Moreover, the current study assumes that mis-reporting occurred equally amongst all eating occasions. This limitation has been reported by previous literature as an important methodological limitation of chrononutrition (Fayet-Moore et al., 2017); in fact further investigation would be warranted to assess the effect of differential misreporting on epidemiological studies in chrono-nutrition in order to suggest possible corrections, e.g. for differential under-reporting at different times of the day (e.g. main meals vs. snack times).

CONCLUSION

- 192 In summary, our study indicates that foods consumed in the evening/night time tend to be highly processed,
- 193 easily accessible, and rich in added sugar or saturated fat, whatever the diabetic status. . Individuals
- 194 with undiagnosed diabetes are more likely to consume specific unhealthy foods at night. The survey
- 195 cross-sectional nature warrants further investigations by longitudinal cohort studies to establish the causal
- 196 relation between time of eating of unhealthy foods and diabetes.

DISCLOSURE/CONFLICT-OF-INTEREST STATEMENT

197 The authors declare that they have no competing interests.

AUTHOR CONTRIBUTIONS

- 198 CW, SA, and LP: designed research and had primary responsibility for final content; CW and LP performed
- 199 statistical analysis; and all authors: wrote the manuscript, read and approved the final manuscript.

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AVAILABILITY OF DATA AND MATERIALS

- 202 Original data used in this study can be accessed upon request to the UK Data Service (https://www.
- 203 ukdataservice.ac.uk) for academic usage (Study Number: 6533).

TABLES

Table 1. The numbers of food recordings contributed to the total calories consumed by adults in the UK adults. (NDNS RP 2008-2017).

Pasta & Rice and other cereals 18353 3512069.99 2.45% 7.36% 7.36	
TTT 1. TO 1	
White Bread 18434 3245641.19 2.46% 6.80% 14.1	
Chips, fried and roast potatoes and potato products 6749 1884058.68 0.90% 3.95% 18.1	2%
Cakes, buns, sweet pastries, fruit pies 7806 1710594.27 1.04% 3.59% 21.7	
Vegetable (not raw) 51317 1665474.02 6.85% 3.49% 25.1	9%
Biscuits 13200 1662598.06 1.76% 3.49% 28.6	3%
Fruit 33903 1641675.02 4.53% 3.44% 32.1	2%
Miscellaneous unclassified foods 48597 1639024.81 6.49% 3.44% 35.5	5%
Chicken/turkey 8863 1617820.30 1.18% 3.39% 38.9	5%
Cheese 10983 1492015.32 1.47% 3.13% 42.0	7%
Beer lager 8199 1484001.20 1.09% 3.11% 45.1	9%
Semi-skimmed milk 57611 1302649.72 7.69% 2.73% 47.9	2%
Potatos other (in salads and dishes) 10113 1291447.61 1.35% 2.71% 50.6	2%
Fat spreads 37960 1215278.60 5.07% 2.55% 53.1	7%
Beef 4987 1124560.42 0.67% 2.36% 55.5	3%
High fiber breakfast cereals 8215 1072813.73 1.10% 2.25% 57.7	3%
Whole meal bread 7193 1070695.89 0.96% 2.24% 60.0	2%
Chocolate 6495 1046112.65 0.87% 2.19% 62.2	2%
Wine 6967 1027792.96 0.93% 2.15% 64.3	7%
Brown, granary and wheatgerm bread 6183 1009074.95 0.83% 2.12% 66.4	3%
Butter 10203 965901.11 1.36% 2.02% 68.5	1%
Eggs 7554 964769.19 1.01% 2.02% 70.5	3%
Soft drinks not diet 11387 940516.516 1.52% 1.97% 72.5)%
Reduced fat spreads 12620 848834.89 1.68% 1.78% 74.2	3%
Crisps and savoury snacks 5664 835671.58 0.76% 1.75% 76.0	1%
Sausages 3025 775004.13 0.40% 1.62% 77.6	5%
Meat pastries 1979 744639.89 0.26% 1.56% 79.2	2%
Bacon and ham 8467 738727.49 1.13% 1.55% 80.7	7%
Yogurt 6776 665484.55 0.90% 1.40% 82.1	5%
Low-fiber breakfast cereals 4303 560296.32 0.57% 1.17% 83.3	1%
Nuts and seeds 6259 559873.88 0.84% 1.17% 84.5	1%
Oily fish 2610 550425.36 0.35% 1.15% 85.6	7%
Whole Milk 13628 530449.07 1.82% 1.11% 86.7	3%
White fish, shellfish 1597 498928.82 0.21% 1.05% 87.8	2%
Puddings 2291 459784.62 0.31% 0.96% 88.7	
Other Milk Cream 6605 434239.37 0.88% 0.91% 89.7	
Pork 1832 420503.76 0.24% 0.88% 90.5	3%

NDNS RP: National Diet and Nutrition Survey.

Table 2. Odds ratio (99% confidence intervals) for food groups eaten at night (8 pm - 6 am) vs. earlier time in the day, among total and according to different diabetes status, NDNS RP 2008-2017.

Selected food group	Overall	Healthy	Pre-diabetics	Undiagnosed diabetics	Diabetics
Pudding	1.43 (1.06, 1.94)	1.55 (1.13, 2.15)	0.95 (0.17, 5.20)	1.82 (0.41, 8.03)	0.63 (0.15, 2.66)
Regular soft drink	1.72 (1.44, 2.05)	1.70 (1.41, 2.05)	1.78 (0.90, 3.48)	2.82 (1.24, 6.43)	1.36 (0.59, 3.10)
Sugar confectionery	1.84 (1.31, 2.59)	1.55 (1.08, 2.23)	2.13 (0.34, 13.24)	10.51 (2.35, 47.04)	5.94 (1.86, 19.00)
Chocolate	3.08 (2.62, 3.62)	2.98 (2.51, 3.54)	4.06 (1.98, 8.31)	2.41 (0.88, 6.60)	4.92 (2.38, 10.20)
;!– Spirit	11.33 (8.42, 15.23)	10.99 (7.99, 15.12)	8.53 (2.28, 31.97)	8.75 (3.00, 25.52)	36.58 (7.36, 181.76)
–¿, Beer	7.26 (5.91, 8.92)	7.55 (6.04, 9.43)	4.42 (2.19, 8.95)	8.29 (3.70, 18.56)	5.82 (2.03, 16.68)
Ice cream	2.45 (1.84, 3.25)	2.52 (1.86, 3.41)	3.39 (0.77, 14.89)	1.07 (0.15, 7.77)	1.74 (0.57, 5.32)
Biscuit	1.90 (1.68, 2.16)	1.78 (1.55, 2.05)	3.25 (1.99, 5.28)	2.96 (1.43, 6.10)	2.33 (1.45, 3.77)
Crisp	1.49 (1.22, 1.82)	1.49 (1.21, 1.85)	2.21 (0.90. 5.41)	1.59 (0.43, 5.95)	0.89 (0.34, 2.33)

Logistic regression models with GEE were adjusted for age, sex, body mass index, and social-economic levels. NDNS RP: National Diet and Nutrition Survey Rolling Programme.

REFERENCES

Almoosawi, S., Cole, D., Nicholson, S., Bayes, I., Teucher, B., Bates, B., Mindell, J., Tipping, S., Deverill, 205 C., and Stephen, A. (2014). Biomarkers of diabetes risk in the national diet and nutrition survey rolling 206 programme (2008–2011). J Epidemiol Community Health 68, 51–56.

- 207 Almoosawi, S., Vingeliene, S., Karagounis, L., and Pot, G. (2016). Chrono-nutrition: A review of current
- evidence from observational studies on global trends in time-of-day of energy intake and its association
- 209 with obesity. *Proc Nutr Soc* 75, 487–500.
- 210 Balieiro, L. C. T., Rossato, L. T., Waterhouse, J., Paim, S. L., Mota, M. C., and Crispim, C. A. (2014).
- Nutritional status and eating habits of bus drivers during the day and night. *Chronobiology international*
- 212 31, 1123–1129.
- 213 Barbadoro, P., Santarelli, L., Croce, N., Bracci, M., Vincitorio, D., Prospero, E., and Minelli, A. (2013).
- 214 Rotating shift-work as an independent risk factor for overweight italian workers: A cross-sectional
- study. PLoS One 8.
- 216 Bates, B., Lennox, A., Prentice, A., Bates, C. J., Page, P., Nicholson, S., and Swan, G. (2014). National
- Diet and Nutrition Survey: Results from years 1, 2, 3 and 4 (combined) of the Rolling Programme
- 218 (2008/2009-2011/2012): A survey carried out on behalf of Public Health England and the Food
- 219 Standards Agency.
- 220 Bingham, S. A., Welch, A. A., McTaggart, A., Mulligan, A. A., Runswick, S. A., Luben, R., Oakes, S.,
- Khaw, K. T., Wareham, N., and Day, N. E. (2001). Nutritional methods in the european prospective
- investigation of cancer in norfolk. *Public health nutrition* 4, 847–858.
- Bonnell, E. K., Huggins, C. E., Huggins, C. T., McCaffrey, T. A., Palermo, C., and Bonham, M. P. (2017).
- Influences on dietary choices during day versus night shift in shift workers: A mixed methods study.
- 225 *Nutrients* 9, 193.
- 226 Cade, J. E., Burley, V. J., Greenwood, D. C., Group, U. W. C. S. S., and others (2004). The UK
- women's cohort study: Comparison of vegetarians, fish-eaters and meat-eaters. *Public health nutrition*
- 228 7, 871–878.
- 229 Chapman, A. N., Beh, E. J., and Palla, L. (2017). Application of correspondence analysis to graphically
- investigate associations between foods and eating locations.
- 231 Fayet-Moore, F., McConnell, A., Kim, J., and Mathias, K. C. (2017). Identifying eating occasion-based
- opportunities to improve the overall diets of australian adolescents. *Nutrients* 9. doi:10.3390/nu9060608.
- 233 Glynn, L., Emmett, P., Rogers, I., and Team, A. S. (2005). Food and nutrient intakes of a population sample
- of 7-year-old children in the south-west of england in 1999/2000–what difference does gender make?
- *Journal of Human Nutrition and Dietetics* 18, 7–19.
- 236 Greenacre, M. (2017). Correspondence analysis in practice. New York: Chapman; Hall.
- 237 Haldar, S., Egli, L., De Castro, C. A., Tay, S. L., Koh, M. X. N., Darimont, C., Mace, K., and Henry, C.
- J. (2020). High or low glycemic index (GI) meals at dinner results in greater postprandial glycemia
- compared with breakfast: A randomized controlled trial. BMJ open diabetes research & care 8.
- 240 doi:10.1136/bmjdrc-2019-001099.
- 241 Hassen, W. S., Castetbon, K., Tichit, C., Péneau, S., Nechba, A., Ducrot, P., Lampuré, A., Bellisle, F.,
- 242 Hercberg, S., and Méjean, C. (2018). Energy, nutrient and food content of snacks in french adults.
- 243 Nutrition Journal 17, 33.
- 244 Kassambara, A., and Mundt, F. (2019). Factoextra: Extract and visualize the results of multivariate data
- 245 analyses. Available at: https://CRAN.R-project.org/package=factoextra.

- 246 Kutsuma, A., Nakajima, K., and Suwa, K. (2014). Potential association between breakfast skipping
- and concomitant late-night-dinner eating with metabolic syndrome and proteinuria in the japanese
- population. *Scientifica* 2014.
- 249 Lê, S., Josse, J., and Husson, F. (2008). FactoMineR: A package for multivariate analysis. Journal of
- 250 *Statistical Software* 25, 1–18. doi:10.18637/jss.v025.i01.
- 251 Mattson, M. P., Allison, D. B., Fontana, L., Harvie, M., Longo, V. D., Malaisse, W. J., Mosley, M.,
- Notterpek, L., Ravussin, E., Scheer, F. A., et al. (2014). Meal frequency and timing in health and
- disease. *Proceedings of the National Academy of Sciences* 111, 16647–16653.
- 254 MRC Elsie Widdowson Laboratory, and NatCen Social Research (2018). National diet and nutrition survey
- years 1-8, 2008/09-2015/16 [data collection]. doi:\url{10.5255/ukda-sn-6533-11}.
- 256 Nakajima, K., and Suwa, K. (2015). Association of hyperglycemia in a general japanese population with
- late-night-dinner eating alone, but not breakfast skipping alone. Journal of Diabetes & Metabolic
- 258 *Disorders* 14, 16.
- 259 NatCen Social Research, MRC Elsie Widdowson Laboratory, University College London. Medical School.
- 260 (2018). National diet and nutrition survey years 1-8, 2008/09-2015/16.
- 261 Nitta, A., Imai, S., Kajiyama, S., Miyawaki, T., Matsumoto, S., Ozasa, N., Kajiyama, S., Hashimoto,
- 262 Y., Tanaka, M., and Fukui, M. (2019). Impact of different timing of consuming sweet snack
- on postprandial glucose excursions in healthy women. Diabetes & metabolism 45, 369-374.
- doi:10.1016/j.diabet.2018.10.004.
- Palla, L., Chapman, A., Beh, E., Pot, G., and Almiron-Roig, E. (2020). Where do adolescents eat less-
- 266 healthy foods? Correspondence analysis and logistic regression results from the UK national diet and
- nutrition survey. *Nutrients* 12, 2235.
- 268 Pan, A., Schernhammer, E. S., Sun, Q., and Hu, F. B. (2011). Rotating night shift work and risk of type 2
- 269 diabetes: Two prospective cohort studies in women. *PLoS Med* 8, e1001141.
- 270 Price, G., Paul, A., Cole, T., and Wadsworth, M. J. (1997). Characteristics of the low-energy reporters in a
- longitudinal national dietary survey. *British Journal of Nutrition* 77, 833–851.
- 272 R Core Team (2019). R: A language and environment for statistical computing. Vienna, Austria: R
- Foundation for Statistical Computing Available at: https://www.R-project.org/.
- 274 Roberts, C., Steer, T., Maplethorpe, N., Cox, L., Meadows, S., Nicholson, S., Page, P., and Swan, G. (2018).
- National Diet and Nutrition Survey: Results from years 7 and 8 (combined) of the Rolling Programme
- 276 (2014/2015–2015/2016).
- 277 Rose, D., and Pevalin, D. J. (2005). The national statistics socio-economic classification: Origins,
- 278 *development and use.*
- 279 Sakai, R., Hashimoto, Y., Ushigome, E., Miki, A., Okamura, T., Matsugasumi, M., Fukuda, T., Majima, S.,
- Matsumoto, S., Senmaru, T., et al. (2017). Late-night-dinner is associated with poor glycemic control
- in people with type 2 diabetes: The KAMOGAWA-DM cohort study. *Endocrine journal*, EJ17–0414.
- 282 SAS Institute (2013). SAS 9.4 language reference: concepts. USA: SAS Institute Inc.
- 283 Slowikowski, K. (2019). Ggrepel: Automatically position non-overlapping text labels with 'ggplot2'.
- Available at: https://CRAN.R-project.org/package=ggrepel.

- Wickham, H. (2016). *ggplot2: Elegant graphics for data analysis*. New York: Springer-Verlag Available at: https://ggplot2.tidyverse.org.
- Wolever, T. M., Jenkins, D. J., Ocana, A. M., Rao, V. A., and Collier, G. R. (1988). Second-meal effect: low-
- 288 glycemic-index foods eaten at dinner improve subsequent breakfast glycemic response. *The American*
- 289 *Journal of Clinical Nutrition* 48, 1041–1047. doi:10.1093/ajcn/48.4.1041.
- 290 Xiao, Q., Garaulet, M., and Scheer, F. A. (2019). Meal timing and obesity: Interactions with macronutrient intake and chronotype. *International Journal of Obesity* 43, 1701–1711.
- Yoshida, J., Eguchi, E., Nagaoka, K., Ito, T., and Ogino, K. (2018). Association of night eating habits with metabolic syndrome and its components: A longitudinal study. *BMC Public Health* 18, 1366.