



Fruit and vegetable intake and risk of incident of type 2 diabetes: results from the consortium on health and ageing network of cohorts in Europe and the United States (CHANCES)

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1 Original article

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3 Title: Fruit and vegetable intake and risk of incident of type 2 diabetes: results from the consortium on
4 health and ageing network of cohorts in Europe and the United States (CHANCES)

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- 61 Ethical standard: All of the cohorts obtained ethical approval and written informed consent was
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63 Abstract:

Background There is limited information to support definitive recommendations concerning the role
of diet in the development of type II Diabetes mellitus (T2DM). The results of the latest metaanalyses suggest that an increased consumption of green leafy vegetables may reduce the incidence
of diabetes, with either no association or weak associations demonstrated for total fruit and
vegetable intake. Few studies have, however, focused on older subjects.

69 Methods The relationship between T2DM and fruit and vegetable intake was investigated using data 70 from the NIH-AARP study and the EPIC elderly study. All participants below the age of 50 and/or with 71 a history of cancer, diabetes or coronary heart disease were excluded from the analysis. Multivariate 72 logistic regression analysis was used to calculate the odds ratio of T2DM comparing the highest with 73 the lowest estimated portions of fruit, vegetable, green leafy vegetables and cabbage intake.

Results Comparing people with the highest and the lowest estimated portions of fruit, vegetable or
green leafy vegetable intake indicated no association with the risk of T2DM. However, although the
pooled OR across all studies showed no effect overall, there was significant heterogeneity across
cohorts and independent results from the NIH-AARP study showed that fruit and green leafy
vegetable intake was associated with a reduced risk of T2DM OR 0.95 (95% CI 0.91,0.99) and OR 0.87
(95% CI 0.87,0.90) respectively.

Conclusion Fruit and vegetable intake was not shown to be related to incident T2DM in older
 subjects. Summary analysis also found no associations between green leafy vegetable and cabbage
 intake and the onset of T2DM. Future *dietary pattern* studies may shed light on the origin of the
 heterogeneity across populations.

84

85 Introduction

The chronic hyperglycaemia that characterizes Type 2 diabetes mellitus (T2DM) is caused by 86 impaired insulin secretion or action and results from the interaction between a genetic 87 88 predisposition and environmental risk factors [1]. In 2004, an estimated 3.4 million people died from 89 the consequences of diabetes or pre-diabetes. According to the WHO, this number is rising and will 90 lead to diabetes being the 7th leading cause of death by 2030 [2]. Although the genetic basis of 91 T2DM has yet to be identified, there is strong evidence that modifiable risk factors such as obesity 92 and a sedentary lifestyle are among the non-genetic determinants of the disease [3-6]. However, 93 other than avoidance of obesity, there is limited information for definitive recommendations 94 regarding the role of diet in the development of T2DM [7-9]. The role of fruit and vegetable intake 95 and risk of T2DM is even less recognized, especially with regards to green leafy vegetables, a rich 96 source of polyphenols which are thought to be associated with increased insulin sensitivity [10].

97 The results of a recent meta-analysis suggests that an increased consumption of fruit and green leafy vegetables may be associated with a significantly reduced risk of T2DM, with no 98 99 association or weak associations demonstrated for total vegetable intake. However, the former 100 observation regarding green leafy vegetables is based on a limited number of studies [11]. 101 Conversely, another more up-to-date meta-analysis reported a dose dependent association between 102 fruit and vegetable intakes separately and a reduced risk of T2DM [12]. An earlier a meta-analysis 103 carried out in 2010 [10] included a sub-analysis using studies with information on green leafy 104 vegetable consumption. The summary estimates showed that greater intake of green leafy 105 vegetables was associated with a 14% reduction in risk of T2DM. Similarly a meta-analysis by Cooper 106 et al [13] also included a sub-analysis of green leafy vegetable intake showing an inverse association 107 with T2DM. Neither study, however, was specifically focussed on older subjects. Therefore, the 108 present study was undertaken to examine the association between T2DM and fruit and vegetables 109 intake, including green leafy vegetables.

110 Methods

111 Study population

- 112 The aim of the Consortium on Health and Ageing Network of Cohorts in Europe and the United
- 113 States (CHANCES) was to combine and integrate prospective cohort studies to produce, improve and
- 114 clarify the evidence on ageing-related health characteristics and risk factors for chronic diseases in
- the elderly, and their socio-economic implications (www.chancesfp7.eu). Detailed characteristics of
- the cohorts have previously been described [14]. All variables used in the analyses from different
- 117 cohorts were harmonised according to pre-agreed CHANCES data harmonisation rules. All of the
- 118 cohorts obtained ethical approval and written informed consent from all participants.
- 119 Participants, aged 50 years and above, were included from the European Prospective Investigation
- 120 into Cancer and Nutrition elderly study (EPIC Elderly) [15] including Spain, Greece, The Netherlands,
- and Sweden (EPIC was treated as 4 different cohorts in the analysis); and the National Institutes of
- 122 Health (NIH)-AARP Diet and Health Study United States [16].

123 Exclusions

- 124 Prior to the analysis, participants at baseline with missing information on chronic diseases
- 125 (cardiovascular disease, diabetes, and cancer), below 50 years of age, missing or unrealistic
- information on body mass index (BMI) [if BMI >60 kg/m² or <10 kg/m²] and with extreme energy
- 127 intake were excluded (applying the cohort specific definitions).

128 Exposure

Habitual dietary intakes were assessed through compatible methods including food frequency
questionnaires (FFQ) and, in some centers within the EPIC elderly study, records of intake over seven
or 14 days that had been developed and validated within each center. In addition, a computerized
instrument for recall of dietary intake over 24 hours was developed to collect information from a

across countries [17]. The number of FFQ items differed across cohorts. The number of FFQ items
used in EPIC elderly was 200 compared to 124 items used in (NIH)-AARP and were both selfreported. (NIH)-AARP Data were thus harmonized across cohorts regarding definitions of food
groups and nutrient units [18]. Fruit and vegetable intakes were calculated in terms of portions per
day (1 portion = 80g). Green leafy vegetable and cabbage, which were less frequently consumed,
were calculated in portions per week (1 portion = 80g).

stratified random sample of the aggregate cohort. The aim was to calibrate the measurements

140 Outcome

133

141 Information on Incident T2DM was collected through self-administered questionnaires or in

142 interviews. The diagnosis of diabetes after the age of 50 was anticipated to be T2DM, as type 1

diabetes usually develops before the age of 40 [19]. All cohorts included in this analysis did not

144 distinguish between type 1 and type 2 diabetes, except for EPIC Elderly Greece.

145 Covariates

146 Model A of the analysis was adjusted for age and sex. Model B was adjusted for age, sex, BMI kg/m²;

underweight (<18.5), normal (\geq 18.5–<25), overweight (\geq 25–<30), moderately obese (\geq 30–<35) and

severely obese (≥35); habitual vigorous physical activity (yes/no) (defined as vigorous exercise at

least once per week); energy intake (Kcal); alcohol consumption [Light = men (>0g & <40g daily),

women (>0g & <20g daily); moderate = men (\geq 40g & <60g daily), women (\geq 20g & <40g daily); and

heavy = men (≥60g daily), women (≥40g daily)]; education (primary or less, more than primary,

152 college or university); and smoking (never, former, current) in all cohorts.

153 Statistical analysis

All analyses were carried out using STATA IC V.11.2 (Stata- Corp, Texas, USA) code available upon
 request. Multivariate logistic regression analysis was used to calculate the odds ratio (OR) of T2DM
 and 95 % confidence intervals (CI) comparing the highest with the lowest estimated intakes of fruit,

157 vegetable, green leafy vegetables and cabbage. This type of analysis was used as the majority of the 158 cohorts had no precise date of diagnosis during follow-up; hence cox modelling/time to event was 159 not ideal. This analysis was conducted in two stages: deriving first the study-specific estimates and 160 then a combined overall estimate; thereafter it was also stratified by categories of intake per day 161 and by total intake of each of fruit, vegetable, green leafy vegetable and cabbage. Categories were 162 developed to maintain consistency across cohorts and so that comparisons could be easily made. 163 Categories for fruit and vegetables were <1.5, 1.5-2.4, 2.5-3.9 and \geq 4 portions per day. For green 164 leafy vegetables and cabbage, the categories were <1.5, 1.5-2.4, 2.5-3.9 and \geq 4 portions per week. 165 We computed both fixed effects models, and random effects models using the DerSimonian-Laird 166 method [20]. Due to substantial heterogeneity across cohort results as assessed with I²- and Q-167 statistics, random effects estimates are reported as the main results, since random effects models 168 allow for variability of effects across individual studies.

169 Results

170 The number of diabetes cases at follow up across the cohorts was as follows (data not shown): NIH-171 AAPP: 22,782; EPIC Elderly All: 1567; EPIC Elderly Spain: 138; EPIC Elderly Greece: 1077; EPIC Elderly 172 Netherlands: 234; and EPIC Elderly Sweden: 118. The characteristics of subjects in each of the 173 cohorts at baseline are presented in Table 1. EPIC Elderly Spain had a higher proportion of 174 individuals in the overweight BMI category, as well as in the moderately obese category. EPIC Elderly Greece, however, had the highest proportion of individuals in the severely obese category. Although 175 176 the energy intakes (Kcal) were similar across the cohorts, EPIC Elderly Sweden had the lowest 177 intakes. EPIC Elderly Spain had the lowest number of individuals who engaged in vigorous physical 178 activity, while EPIC Elderly Netherlands had the highest proportion of individuals who said they did 179 vigorous activity. 180 The highest proportion of individuals who drank heavily were those of the EPIC Elderly Netherlands 181 cohort. They were also mostly women, with men only making up 5% of the cohort participants. The 182 NIH-AARP study had more highly educated subjects, and the highest number of former smokers. 183 EPIC Elderly Greece had the highest proportion of current smokers. 184 Intakes of fruit and vegetables, which were calculated in portions per day (for fruit and vegetable) or 185 per week (green leafy vegetables and cabbage) are also shown in Table 1. Intakes varied between 186 cohorts especially between subgroups of vegetables. For example, intakes of cabbage were lowest in 187 the EPIC Elderly Sweden and Spain cohorts. EPIC Elderly Greece had the highest intakes across all 188 four categories, across all cohorts, whereas EPIC Elderly Sweden had lowest number of individuals in all four categories. 189 190 Median intakes and ORs (95% CI) for T2DM are presented in (Table 2) and (Table 3) for categories as 191 well as total intake per day (1 portion = 80g). Compared with the lowest category of intake, the 192 multivariate adjusted OR (Model B) of T2DM across categories of fruit showed a slightly reduced risk 193 of T2DM in the NIH-AARP study; OR: 0.95 (95%CI 0.91-0.99). This, however, was not the case in the

194 EPIC Elderly cohorts where no significant associations were found; for example, EPIC Elderly (all), OR: 195 1.01 (95%CI 0.80-1.28). Figure 1 shows the overall pooled multivariate odds ratio for T2DM 196 comparing the highest with the lowest fruit intakes across the NIH-AARP & EPIC Elderly cohorts. The 197 results show no overall association with the risk of T2DM, OR: 1.00 (95%CI 0.83-1.19). Across 198 categories of vegetable intake, there was no association with risk of T2DM across EPIC Elderly (all 199 and separately) after adjustments were made in Model B. A reduced risk of T2DM, comparing the 200 highest to the lowest category of vegetable intake, was apparent in NIH-AARP, OR: 0.92 (95%CI 0.87-201 0.97). In the Spanish and Greek EPIC Elderly cohorts there were non-significant increases in risk of 202 T2DM, OR: 1.42 (95%CI 0.78-2.58) and OR: 2.15 (95%CI 0.93-5.03), respectively. Figure 2 shows the 203 pooled analysis for vegetable intake and T2DM risk. The pooled OR in Model B was 1.13 (95%CI 0.77-204 1.64) indicating no overall association between vegetable intake and incident T2DM. 205 In the NIH-AARP cohort, green leafy vegetable intake was associated with a reduced risk of T2DM 206 which retained its significance in Model B, OR: 0.87 (95%CI 0.84-0.90). However the trends in the 207 EPIC Elderly cohorts were in the opposite direction, with an increase in the odds of developing T2DM 208 in those with the highest intakes of green leafy vegetables; EPIC Elderly All, OR: 1.23(95%CI 1.01-209 1.50), and EPIC Elderly Greece, OR: 1.52 (1.13-2.04). Nevertheless, the pooled analysis, shown in 210 Figure 3, indicated no overall association between intake of green leafy vegetables and T2DM, OR: 211 1.08 (0.80, 1.46). Finally, when compared to the lowest category of intake, those with highest 212 cabbage intakes had a reduced risk of T2DM across the EPIC Elderly Netherlands cohort after 213 adjustments were made in Model B, OR: 0.61(95%CI 0.35-1.05), though the Confidence Limits could 214 not exclude the null value. In the analysis using the NIH-AARP study, there were also associations 215 found between cabbage intakes and incident T2DM, however these indicated a small increased risk 216 for T2DM, OR: 1.07(0.94-1.21). Thus overall, no association was found between cabbage intake and 217 incident T2DM (Figure 4), OR: 1.03 (95%CI 0.90, 1.18).

218 Discussion

219 Associations found between intakes of fruits, vegetables, green leafy vegetables and cabbage and 220 incident T2DM varied, as they showed both a reduced risk of T2DM as well as an increased risk 221 across these CHANCES cohorts. Nevertheless, although there was heterogeneity between cohorts, 222 the overall pooled results using multiple cohorts from different countries showed no association 223 with risk of incident T2DM. Being so large, the NIH AARP study has a major impact on our pooled 224 results so in a separate sensitivity analysis we pooled results for all EPIC Elderly cohorts excluding 225 NIH-AARP, which offered the following results per portion: for fruits OR: 1.07 (95%CI 0.77,1.49); 226 vegetables OR 1.49 (95%CI 0.94, 2.36); green leafy vegetables OR: 1.23 (95%CI 0.93, 1.62) and 227 cabbage OR: 0.90 (95%CI 0.66, 1.23), re-affirming the null associations.

228 Similar results have been shown in two meta-analyses [10, 21]. The systematic review by 229 Hamer & Chaida (2007) also included studies measuring antioxidant intake and incidence of T2DM in 230 a separate meta-analysis. The relative risk of T2DM from consuming five or more servings of fruit 231 and vegetables a day was 0.96 (95% CI, 0.79–1.17, P=0.96), and 1.01 (0.88–1.15, P=0.88) for three or 232 more servings of fruit, and 0.97 (0.86–1.10, P=0.59) for three or more servings of vegetables. The 233 authors concluded that the consumption of three or more servings a day of fruit or vegetables is not 234 associated with a reduction in the risk of T2DM. This was similar to the results by Wu et al (2015) 235 which showed that total fruit and vegetable consumption was not significantly associated with risk 236 of T2DM. However, significant heterogeneity was shown for the combined effects of fruit and 237 vegetables intake in the review by Hamer & Chaida (2007) [10, 21]. This was mostly due to the 238 substantially lower risk estimate among women reported by the study by Ford and Mokdad (2001) 239 [22]. Furthermore, showing somewhat different results, a meta-analysis carried out by Carter et al 240 (2010) included six cohort studies, four of which included information on green leafy vegetable 241 consumption. The pooled estimates showed no significant reduced risk from increasing the 242 consumption of vegetables, fruit, or fruit and vegetables combined, results which accord with those 243 in our current study. Nevertheless, the summary estimates from only four studies which assessed

green leafy vegetable consumption showed that greater intake of green leafy vegetables was
associated with a 14% reduction in risk of T2DM (hazard ratio 0.86, 95% confidence interval 0.77 to
0.97). A similar reduced risk for green leafy vegetables was also noted in two recent meta-analysis by
Li et al, 2014 [11] and Cooper et al, 2015 [23]. However, most of the studies included in the metaanalysis included females only (4/6) and therefore the results may not be generalizable to a wider
population.

250 Several possible mechanisms have been proposed to explain the potential associations between 251 consuming more fruits and vegetables and green leafy vegetables in the diet, and the incidence of 252 T2DM. Fruit and vegetables are rich in fibre, which has been shown to improve insulin sensitivity and 253 insulin secretion [24], though not all studies have found consistent associations with risk of T2DM 254 [25]. On the other hand, many fruits are rich sources of fructose and fructose metabolism may 255 decease insulin sensitivity and increase risk factors for metabolic syndrome and T2DM [26]. 256 Increased intakes of fruit and vegetables have been shown to be inversely associated with obesity 257 [27], which in turn is one of the most established risk factors for T2DM development [28]. The 258 consumption of sugar sweetened fruit juices has also been positively associated with T2DM [29]. 259 Green leafy vegetables confer antioxidant properties, which may mitigate T2DM risk through their 260 high concentrations of β carotene, polyphenols and vitamin C [30, 31]. Additionally, green leafy 261 vegetables could reduce the risk of T2DM due to their magnesium content, which has been shown to 262 play a role in glucose control and improving insulin sensitivity [32]. Furthermore they are particularly 263 rich in inorganic nitrate [33] which has been linked to improvement in reaction time in individuals 264 with T2DM [34]. Thus these various putative mechanisms do not point consistently towards a single 265 direction of effect for fruits and vegetable, making the inconsistent findings from observational 266 studies, in which dose and pattern of consumption are recorded with variable precision, hardly 267 surprising. It is also possible that other specific categories of fruit and vegetables are more closely 268 associated with diabetes risk than overall fruit and vegetable intake, however we were not able to 269 assess this in the current analysis. Intakes of fruit and vegetables are highly correlated with other

270 lifestyle and dietary factors, and so it is difficult to isolate the effect of these intakes on T2DM 271 independent of other factors. Consequently, when interpreting such disparate results, attempts 272 must be made to control for some of the important confounders across the cohorts. 273 Our study has specific strengths and limitations. The main strength was the ability to compare 274 cohorts from different countries which have harmonised the vast majority of variables using 275 individual participant data. However, high levels of heterogeneity were found for the leafy green 276 vegetable analysis (I²=79.3%, p=0.002) and differences in the classification of leafy green vegetables 277 may exist between cohorts. Although all data were harmonised based on agreed rules 278 (www.chancesfp7.eu; [16]), the data from the different cohorts are not perfectly comparable, due to 279 differences in study design and data collection procedures, with the potential for residual 280 inconsistencies in variable definitions. Although we made strenuous effort at harmonisation, the 281 dietary assessment methods used in these studies differed with, for example, the total number of 282 FFQ items differing across the cohorts and with EPIC elderly using more than one method (FFQ/24 283 hour diet recall). This may be a possible explanation for differences found across the cohorts. 284 Similarly, the strengths of the meta-analysis may also be weaknesses where the possibility of the 285 exposure is still heterogeneous for the same reason mentioned above. Individual study odds ratios 286 are presented in Figures 1-4 and show the effects that each study has on the pooled effect estimate. 287 Additionally, under-reporting and selective recall (of healthier foods) can be a problem with 288 unpredictable consequences since dietary constituents are not consumed in isolation. Although we 289 adjusted for several pertinent confounders, residual confounding from unmeasured risk factors 290 cannot be ruled out. We were unable, for example, to analyse dietary patterns and had this been 291 possible it may have shed additional light on the heterogeneity across cohorts, as in some countries 292 the consumption of vegetables by older people correlates highly with intakes of red meat [35] and 293 intakes of meat may be associated with diabetes risk [36]. A further consideration, which was not 294 possible to explore in this study, is the impact of different cooking methods and of the ways fruits

and vegetables are incorporated into meals, and the impact of both on overall micronutrient content[37].

297 Imprecision arising from a single measurement of diet at baseline may also have introduced 298 some bias into this study, though classically this is often assumed to be towards the null [38]. In 299 addition to this, lack of corroboration that the outcome used in this analysis is T2DM, which was an 300 assumption made based on self-reported age of diagnosis, is a limitation of the study, though we do 301 not believe that the precision of outcome verification should be differentially associated with the 302 accuracy of any particular nutrient intake. Furthermore, the risk of under-ascertainment of diabetes 303 might be greater in people who don't visit their doctor very often and these are likely to be the 304 people on healthier diets. This would however not be an explanation of our lack of finding an inverse 305 association. Finally, although having a precise date of diagnosis for the cases ascertained in these 306 CHANCES cohorts would have been preferable, the essentially null findings suggest that a time-to-307 event analysis may not have been particularly illuminating. 308 In summary, while there was some notable heterogeneity across cohorts, this study suggests

that in older subjects there was no overall association between fruit, vegetable, green leafy
vegetable, or cabbage and incident T2DM. Further studies are needed to assess these effects on
T2DM risk in older people.

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Table 1. Baseline characteristics of CHANCES	partici	pants
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Baseline variables	CHANCES cohort					
	NIH-AARP	EPIC Elderly All	EPIC Elderly Spain	EPIC Elderly Greece	EPIC Elderly The Netherlands	EPIC Elderly Sweden
Total N	401909	20629	4309	7567	5786	2967
Mean follow up time of study (years)	10.6	11.8	13.1	10.0	12.6	13.3
Age at baseline ¹	62 ± 5	64 ± 4	63 ± 2	67 ± 5	64 ± 3	60 ± 1
BMI (Kg/m ²) ²						
Underweight	4,343 (1)	134 (0.65)	4 (0.09)	29 90.380	69 (1)	32 (1)
Normal	145,197 (37)	5,650 (27)	496(12)	1,292 (170	2,518 (44)	1,344 (45)
Overweight	166,893 (43)	,850 (43)	2,041 (47)	3,184 (42)	2,412 (42)	1,213 (41)
Modestly obese	56,134 (14)	4,531 (22)	1,367 (32)	2,257 (30)	639 (11)	268 (9)
Severely obese	20,096 (5)	1,464 (7)	401 (9)	805 (11)	148 (3)	110 (4)
Vigorous Physical activity ²	186,334 (46)	5,080 (29)	222 (5)	1,603 (22)	3,255 (58)	na
Energy intake in Kcal ¹	1,822 ± 651	1,835 ± 583	2,036 ± 674	1,842 ± 586	1,772 ± 429	1,647 ± 600
Alcohol consumption/day ²						
0	88,022 (22)	5,771 (28)	1,722 (40)	2,479 (33)	1,197 (21)	373 (13)
1	274,779 (68)	13,031 (63)	1,987 (46)	4,690 (62)	3,701 (65)	2,593 (87)
2	19,638 (5)	1,257 (6)	344 (8)	243 (3)	669 (12)	1 (0.03)
3	19,470 (5)	570 (3)	256 (6)	155 (2)	159 (30)	-
Sex ²						
Male	231,259 (58)	6,394 (31)	1,858 (43)	2,873 (38)	263 (5)	1,400 (47)
Education ²						
Primary or less	2,592 (0.6)	14,071 (68)	3,674 (85)	6,883 (91)	1,896 (33)	1,618 (55)
More than primary	95,522 (24)	4,927 (24)	308 (7)	417 (6)	3,228 (56)	974 (33)
College or University	293,119 (73)	1,523 (7)	277 (6)	238 (3)	655 (11)	353 (12)
Smoking status ²						
Never	147,429 (37)	12,745 (62)	2,896 (67)	5,269 (70)	2,778 (48)	1,802 (61)
Former	190,969 (48)	4,475 (22)	676 (16)	1,230 (16)	1,960 (34)	609 (21)
Current	48,597 (12)	3,126 (15)	734 (17)	857 (11)	1,047 (18)	488 (16)
Fruits and Vegetables ³						
Fruits p/day	3.7 (2.1-5.9)	3.2 (1.9-4.7)	3.6 (2.3-5.7)	4.0 (2.9-5.3)	2.9 (1.6-3.9)	1.7 (0.9-2.9)
Vegetables p/day	3.2 (2.1-4.6)	2.4 (1.4-4.5)	2.5 (1.6-3.8)	4.8 (3.7-6.2)	1.6 (1.2-2.0)	0.7 (0.4-1.4)
Green leafy vegetables p/w	2.3 (0.9-4.8)	2.7 (1.2-5.0)	3.9 (1.7-7.7)	4.1 (2.5-6.2)	2.4 (1.4-3.5)	0.03 (0.01-0.3)
Cabbage p/w	0.3 (0.08-0.6)	1.2 (0.3-2.4)	0.1 (0.0-0.8)	1.9 (0.9-2.8)	1.7 (1.0-2.7)	0.2 (0.02-0.37)

Abbreviations: N: number, BMI: body mass index, kcal: kilocalorie, na: not available, p: portion, w: week. Exclusions: Age<50, History of Coronary Heart Disease, Cancer and Diabetes. Alcohol consumption/day= 0:nondrinker, 1: Light = men (>0g & <40g daily), women (>0g & <20g daily); 2: Moderate = men (≥40g & <60g daily), women (≥20g & <40g daily); 3: Heavy = men (≥60g daily), women (≥40g daily). 1Mean and standard deviation

²Number and percentage

³Median and Interquartile range

	Portions/day of fruit intake					Total intake	Portions/day of vegetable intake			Р	Total intake	
-	<1.5	1.5-2.4	2.5-3.9	≥4		1 portion/day	<1.5	1.5-2.4	2.5-3.9	≥4		
NIH-AARP												
Median	0.82	1.99	3.24	7.73			1.04	2.02	3.20	6.41		
Model Aª: OR (95% CI)	1.00 (Ref)	0.93(0.88-0.97)	0.86(0.82-0.89)	0.86(0.83-0.89)	<0.0 1	1.00(1.00-1.00)	1.00 (Ref)	0.91(0.87-0.95)	0.89(0.85-0.93)	0.98(0.93-1.02)	0.03	1.02(1.01-1.02)
Model B ^b : OR (95% CI)	1.00 (Ref)	0.96(0.91-1.02)	0.95(0.91-0.99)	0.95(0.91-0.99)	0.04	1.00(0.99-1.01)	1.00 (Ref)	0.92(0.87-0.97)	0.88(0.84-0.94)	0.92(0.87-0.97)	0.14	1.00(0.99-1.01)
EPIC Elderly (All)												
Median	0.87	1.89	3.18	5.3			0.97	1.9	3.18	5.5		
Model A ^a : OR (95% CI)	1.00 (Ref)	0.86(0.69-1.07)	0.96(0.79-1.16)	1.01(0.83-1.22)	0.24	1.01(0.98-1.04)	1.00 (Ref)	0.98(0.79-1.21)	1.19(0.95-1.51)	1.23(0.97-1.56)	0.05	1.03(0.99-1.06)
Model B ^b : OR (95% CI)	1.00 (Ref)	0.95(0.73-1.23)	1.01(0.80-1.28)	1.01(0.80-1.28)	0.88	1.00(0.97-1.03)	1.00 (Ref)	0.99(0.79-1.26)	1.11(0.86-1.45)	1.05(0.79-1.37)	0.72	0.99(0.96-1.03)
EPIC Elderly Spain												
Median	0.57	1.94	3.21	5.98			1.05	1.98	3.11	5.03		
Model A ^a : OR (95% CI)	1.00 (Ref)	1.59(0.82-3.13)	1.15(0.61-2.18)	1.47(0.82-2.63)	0.36	1.02(0.96-1.08)	1.00 (Ref)	1.76(1.04-2.96)	1.52(0.90-2.56)	1.35(0.76-2.37)	0.76	0.99(0.91-1.99)
Model B ^b : OR (95% CI)	1.00 (Ref)	1.83(0.91-3.67)	1.28(0.66-2.54)	1.75(0.94-3.26)	0.17	1.04(0.98-1.0)	1.00 (Ref)	1.89(1.10-3.26)	1.66(0.96-2.87)	1.42(0.78-2.58)	0.72	0.99(0.90-1.00)
EPIC Elderly Greece												
Median	1.06	2.08	3.28	5.29			1.15	2.12	3.39	5.61		
Model A ^a : OR (95% CI)	1.00 (Ref)	1.04(0.72-1.49)	1.08(0.78-1.49)	1.13(0.83-1.56)	0.24	1.01(0.98-1.04)	1.00 (Ref)	2.14(0.88-5.16)	2.62(1.14-6.07)	2.73(1.19-6.27)	0.05	1.03(0.99-1.07)
Model B ^b : OR (95% CI)	1.00 (Ref)	1.12(0.77-1.64)	1.09(0.77-1.54)	1.09(0.77-1.55)	0.88	1.00(0.96-1.04)	1.00 (Ref)	1.96(0.81-4.77)	2.29(0.99-5.36)	2.15(0.93-5.03)	0.72	0.99(0.95-1.04)
EPIC Elderly												
Netherlands												
Median	0.96	1.73	3.14	4.80			1.17	1.86	2.81	4.33		
Model A ^a : OR (95% CI)	1.00 (Ref)	0.59(0.39-0.88)	0.84(0.58-1.21)	0.73(0.49-1.09)	0.63	0.96(0.89-1.04)	1.00 (Ref)	0.80(0.6-1.06)	0.87(0.56-1.37)	0.71(0.09-5.33)	0.29	0.97(0.79-1.19)
Model B ^b : OR (95% CI)	1.00 (Ref)	0.57(0.37-0.88)	0.89(0.61-1.32)	0.73(0.48-1.12)	0.74	0.96(0.88-1.04)	1.00 (Ref)	0.72(0.54-0.98)	0.73(0.46-1.17)	0.92(0.12-7.12)	0.09	0.90(0.73-1.12)
EPIC Elderly Sweden												
Median	0.81	1.89	3.08	4.50			0.58	1.89	2.96	4.67		
Model A ^a : OR (95% CI)	1.00 (Ref)	0.78(0.49-1.27)	0.88(0.52-1.49)	1.21(0.68-2.16)	0.59	1.07(0.96-1.19)	1.00 (Ref)	0.75(0.41-1.39)	2.17(1.17-4.00)	0.53(0.07-3.92)	0.42	1.11(0.93-1.32)
Model B ^b : OR (95% CI)	1.00 (Ref)	0.72(0.43-1.19)	0.87(0.50-1.52)	1.16(0.62-2.15)	0.67	1.07(0.95-1.20)	1.00 (Ref)	0.68(0.35-1.29)	2.17(1.14-4.13)	0.46(0.06-3.46)	0.59	1.09(0.91-1.31)

Table 2. Association between Diabetes, fruits and vegetables in CHANCES participants

Abbreviations: g: grams; OR: odds ratio; CI: confidence Intervals. ^aModel A: adjusted for sex and age. ^bModel B: adjusted for age, sex, BMI, physical activity, energy intake, alcohol consumption, education and smoking. *P* is for trend across categories.

	Portions/week of leafy green vegetables intake				Р	Total intake	Portions/week of cabbage intake				Р	Total intake
	<1.5	1.5-2.4	2.5-3.9	≥4		1 portion/day	<1.5	1.5-2.4	2.5-3.9	≥4		
NIH-AARP												
Median	0.65	1.98	3.10	8.06			0.32	1.63	3.90	9.79		
Model A ^a : OR (95% CI)	1.00 (Ref)	0.86(0.83-0.89)	0.81(0.77-0.85)	0.82(0.79-0.85)	< 0.01	0.98(0.98-0.98)	1.00 (Ref)	1.09(1.05-1.15)	1.24(1.16-1.33)	1.19(1.06-1.33)	< 0.01	1.04(1.03-1.05)
Model B ^b : OR (95% CI)	1.00 (Ref)	0.90(0.86-0.94)	0.89(0.85-0.94)	0.87(0.84-0.90)	< 0.01	0.98(0.98-0.99)	1.00 (Ref)	1.06(1.01-1.12)	1.09(1.00-1.18)	1.07(0.94-1.21)	< 0.01	1.02(1.01-1.03)
EPIC Elderly (All)												
Median	0.32	2.04	3.14	6.18			0.37	1.99	3.07	4.96		
Model A ^a : OR (95% CI)	1.00 (Ref)	1.11(0.89-1.38)	1.28(1.05-1.57)	1.30(1.09-1.59)	< 0.01	1.02(0.99-1.03)	1.00 (Ref)	0.94(0.82-1.09)	1.16(1.00-1.34)	0.98(0.80-1.19)	0.02	1.00(0.97-1.04)
Model B ^b : OR (95% CI)	1.00 (Ref)	1.09(0.87-1.37)	1.25(1.01-1.53)	1.23(1.01-1.50)	0.02	1.00(0.99-1.02)	1.00 (Ref)	0.89(0.76-1.04)	1.09(0.94-1.27)	0.96(0.77-1.19)	0.15	0.99(0.94-1.03)
EPIC Elderly Spain												
Median	0.55	1.93	3.14	7.66			0	2.11	2.93	4.95		
Model A ^a : OR (95% CI)	1.00 (Ref)	1.77(1.01-3.14)	1.28(0.73-2.24)	1.02(0.64-1.62)	0.44	0.98(0.95-1.02)	1.00 (Ref)	0.79(0.38-1.63)	0.96(0.38-2.39)	0.79(0.37-1.73)	0.49	0.94(0.83-1.06)
Model B ^b : OR (95% CI)	1.00 (Ref)	1.74(0.97-3.11)	1.39(0.79-2.47)	1.04(0.64-1.68)	0.52	0.98(0.95-1.01)	1.00 (Ref)	0.86(0.41-1.78)	0.96(0.38-2.40)	0.80(0.37-1.75)	0.54	0.95(0.84-1.07)
EPIC Elderly Greece												
Median	0.87	2.13	3.13	6.18			0.84	2.06	3.06	4.88		
Model A ^a : OR (95% CI)	1.00 (Ref)	1.26(0.92-1.72)	1.57(1.18-2.11)	1.61(1.22-2.12)	< 0.01	1.03(1.01-1.05)	1.00 (Ref)	0.99(0.83-1.18)	1.28(1.09-1.51)	1.14(0.90-1.43)	0.02	1.03(0.99-1.07)
Model B ^b : OR (95% CI)	1.00 (Ref)	1.23(0.89-1.71)	1.55(1.14-2.11)	1.52(1.13-2.04)	0.02	1.02(0.99-1.04)	1.00 (Ref)	0.93(0.77-1.11)	1.21(1.07-1.44)	1.09(0.85-1.41)	0.15	1.02(0.98-1.07)
EPIC Elderly												
Netherlands												
Median	0.97	1.99	3.14	4.95			0.92	1.92	3.08	4.97		
Model A ^a : OR (95% CI)	1.00 (Ref)	0.88(0.61-1.29)	0.99(0.65-1.33)	1.13(0.77-1.68)	0.49	1.05(0.97-1.14)	1.00 (Ref)	0.82(0.59-1.12)	0.74(0.51-1.08)	0.57(0.33-0.97)	0.01	0.86(0.79-0.96)
Model B ^b : OR (95% CI)	1.00 (Ref)	0.89(0.61-1.31)	0.87(0.60-1.26)	1.03(0.69-1.54)	0.86	1.02(0.94-1.10)	1.00 (Ref)	0.81(0.58-1.11)	0.79(0.54-1.16)	0.61(0.35-1.05)	0.04	0.87(0.78-0.97)
EPIC Elderly Sweden												
Median	0.03	1.77	3.52	7.04			0.17	1.88	3.75	5.25		
Model A ^a : OR (95% CI)	1.00 (Ref)	0.88(0.12-6.56)	2.86(0.52-9.60)	-	0.60	1.12(0.89-1.42)	1.00 (Ref)	0.85(0.27-2.75)	1.50(0.68-3.31)	0.67(0.09-4.90)	0.73	1.07(0.92-1.23)
Model B ^b : OR (95% CI)	1.00 (Ref)	0.76(0.09-5.86)	2.98(0.85-10.4)	-	0.67	1.12(0.87-1.43)	1.00 (Ref)	0.77(0.23-2.59)	1.53(0.68-3.48)	0.76(0.10-5.71)	0.69	1.04(0.90-1.21)

Table 3. Association between Diabetes, green leafy vegetables and cabbage in CHANCES participants

Abbreviations: g: grams; OR: odds ratio; CI: confidence Intervals. ^aModel A: adjusted for sex and age. ^bModel B: adjusted for age, sex, BMI, physical activity, energy intake, alcohol consumption, education and smoking. *P* is for trend across categories.







Cabbage intakes (portions/week)	OR (95%CI)	% Weight
Cohort		
NIH-AARP	1.07 (0.94, 1.22)	65.26
EPIC elderly Spain	0.80 (0.37, 1.73)	3.08
EPIC elderly Greece	1.09 (0.85, 1.40)	25.37
EPIC elderly Netherlands	0.61 (0.35, 1.06)	5.84
EPIC elderly Sweden	0.76 (0.10, 5.78)	0.45
Overall (I-squared = 9.1%, p = 0.355)	1.03 (0.90, 1.18)	100.00
NOTE: Weights are from random effects analysis		
.1 1	10	