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SSM-D-18-00856**Fat tax or thin subsidy? How price increases and decreases affect the energy and nutrient content of food and beverage purchases in Great Britain**Laura Cornelsen^a, Mario Mazzocchi^b, Richard Smith^{c,a}^a Faculty of Public Health and Policy, London School of Hygiene and Tropical Medicine, 15-17 Tavistock Place, WC1H 9SH London, United Kingdom.^b Department of Statistical Sciences, University of Bologna, Via Belle Arti 41, Bologna, Italy.^c College of Medicine and Health, University of Exeter, Medical School Building, St Luke's Campus, Magdalen Road, Exeter, EX1 2LU, United Kingdom.Corresponding author: Laura Cornelsen ; email: laura.cornelsen@lshtm.ac.uk**Acknowledgments**

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1 **Fat tax or thin subsidy? How price increases and decreases affect** 2 **the energy and nutrient content of food and beverage purchases** 3 **in Great Britain**

4 5 **1. Introduction**

6 Poor diets are the second leading risk factor for mortality globally, accounting for nearly one
7 in five deaths (GBD 2016 Risk Factors Collaborators). Diets low in whole grains and fruits,
8 and high in sodium, are leading individual dietary risks to mortality. At the same time
9 consuming too much red meat and sugar-sweetened beverages are individual factors seeing
10 the greatest increase in attributable deaths and DALY's since 1990 (GBD 2016 Risk Factors
11 Collaborators).

12 Policies to change food price are increasingly recommended to prevent diet-related diseases
13 at population level, particularly in middle- and high-income countries (WHO, 2018). The
14 underlying idea behind these policies is simple: if unhealthy foods became more expensive
15 (e.g. via taxes), or healthy food became cheaper (e.g. via subsidies), consumers would
16 decrease or increase their purchase, and consumption, accordingly.

17 The reality, however, is more complicated as there are difficulties in measuring the impacts
18 from individual food price changes on total diet as consumers may change their demand for
19 many other foods in response to the policy, negating or exacerbating the intended effect.
20 Thus, the comparative effectiveness of either price increase or decreases in improving the
21 overall quality of diets, at population level, remains unknown.

22 Reviews of prospective modelling studies of taxes on foods, mostly on sugary drinks,
23 generally suggest that a 20% tax can increase the price sufficiently to reduce purchases of the

1 taxed products (Afshin et al., 2017; Backholer et al., 2016; Thow et al., 2014; Wright et al.,
2 2017). Whether this translates to a reduction in obesity prevalence and associated disease is
3 less clear (Bes-Rastrollo et al., 2016; Nakhimovsky et al., 2016). Studies find nutrient-based
4 taxes (e.g. tax on sugar) to be comparatively more effective than product-based taxes
5 (Harding & Lovenheim, 2017). However, from policy perspective these are also more
6 complex to implement. Recent studies from countries in which taxes have been levied on
7 unhealthy foods or beverages confirm a small reduction in their consumption (Bíró, 2015;
8 Bødker et al., 2015; Colchero et al., 2017; Silver et al., 2017). Small positive health effects
9 have only been shown through modelling actual changes in purchases for the Danish tax on
10 saturated fats in 2011 (Smed & al., 2016) and for the tax on sugary drinks in Mexico
11 (Barrientos-Gutierrez et al., 2017).

12 For consumers, subsidies on healthy foods are understandably more acceptable than taxes on
13 unhealthy foods (Mazzocchi et al., 2014) prompting a discussion on whether price subsidies
14 (on their own or together with taxes) are a more appropriate solution (Caraher & Cowburn,
15 2015; Niebylski et al., 2015). Experimental studies demonstrate the effectiveness of subsidies
16 to improve diets in specific contexts (e.g. among limited population, geographical areas or
17 implemented in selected outlets), often in parallel with other interventions, such as provision
18 of information (Afshin et al., 2017; Epstein et al., 2012). Wider subsidy programs covering
19 large population shares (and evaluations of their effectiveness) in the context of improving
20 dietary health, rather than addressing basic food security needs, are rare. Most evidence
21 comes from the recipients of Supplemental Nutrition Assistance Programme (SNAP) in the
22 US, showing an increase in fruit and vegetable consumption when incentivised by a price
23 subsidy (Chang et al., 2015; Kaushal & Muchomba, 2015; Klerman et al., 2014).

24 This study contributes to the evidence on the impact of price changes on the nutritional
25 quality of diets by exploiting a unique data-set on food purchases. Home-scan data from

1 Kantar Worldpanel, for a large and representative sample of more than 26,000 British
2 households, opens the way to the estimation of the potential outcomes of price increases and
3 decreases on energy, and nutrients purchased from a range of healthier and less healthy foods
4 and beverages. These cover the whole food basket and thus the full range of own- and cross-
5 price effects.

6 The novelty in our approach lies in two aspects. First, we measure the impact of price
7 changes on purchases of energy and nutrients and thereby explicitly measure changes in
8 dietary quality while maintaining modelling choices among food groups, rather than
9 nutrients. This is an important distinction because when people shop they shop for specific
10 foods rather than energy and nutrients. By categorising all foods based on their relative
11 healthiness using a nutrient profile model used in Great Britain since 2004 (UK Department
12 of Health, 2011) we also indicate food groups for which price changes may lead to the largest
13 gains, as well as potential trade-offs. For example, an increase in the price of foods high in
14 sugar might have a knock-on effect on fibre consumption if foods have relatively high levels
15 of both, as in many breakfast cereals.

16 Second, given the highly disaggregated nature of the data over time, our demand model
17 specification is generalised to allow elasticities to depend on the difference with previous
18 shopping prices. Such generalisation is driven by long-standing empirical marketing research
19 (including in food demand) that the same consumer might react with different intensity to
20 price increases and decreases (Bell & Lattin, 2000; Han et al., 2001; Juhl & Jensen, 2014;
21 Kalyanaram & Winer, 1995; Krishnamurthi et al., 1992; Mazumdar et al., 2005; Taludkar &
22 Lindsey, 2013). While it is difficult to summarise the magnitude of such asymmetry in price
23 response, as the papers look at very different products using different models, the conclusions
24 support the existence of asymmetry in price response to increasing and decreasing prices.

1 The most credited explanation for asymmetry in demand is based on the reference-price
 2 concept; where consumers compare current prices with a (reference) price that has been
 3 formed from past shopping experiences. If the current price differs from this reference price
 4 the consumer experiences gains or losses. This is also consistent with Prospect Theory, which
 5 predicts that consumers react more strongly to losses than to gains (Tversky & Kahneman,
 6 1991), such that the impact of price increases produces a more pronounced effect on
 7 consumption in comparison to an equivalent price decrease. Although traditional demand
 8 models assume symmetric response, and we relax this assumption, the case of symmetric
 9 response remains as an empirical restriction to the model applied.

10

11 2. Methods

12 2.1 Demand model

13 We follow the Almost Ideal Demand System (AIDS) specification with modification to allow
 14 for inclusion of price variation relative to the prices observed in the previous shopping
 15 occasion to identify price increases and price decreases (Cornelsen et al., 2016). This allows
 16 estimating price elasticities of both directions of price change. This specification is consistent
 17 with the concept of reference prices, under the assumption that the consumer reference prices
 18 in current shopping occasion are those of the previous occasion they were in the same shop.
 19 The use of reference price is intuitive as it is reasonable to assume that in a weekly context
 20 shoppers are very likely to consider a reference price in purchase decisions.

21 The demand function for the i -th item is specified as follows:

$$\begin{aligned}
 w_{iht} = & \alpha_i + \sum_{j=1}^n \gamma_{ij} \log p_{jht} + \sum_{j=1}^n \delta_{ij} I_{jht} (\log p_{jht} - \log p_{jh,t-1}) \\
 & - \sum_{j=1}^n \omega_{ij} (1 - I_{jht}) (\log p_{jht} - \log p_{jh,t-1}) + \beta_i \log \left(\frac{x_{ht}}{p_{ht}} \right) + u_{iht} \quad (1)
 \end{aligned}$$

22

1 Where:

2 w_{iht} is expenditure share of group i ($i=1,2,\dots,26$) for household h in week t ($t=1,2,\dots,104$)

3 p_{jht} are prices across food groups j ($j=1,2,\dots,26$) for household h in week t ($t=1,2,\dots,104$)

4 I_{iht} is an indicator function which equals to 1 if $p_{jht} > p_{jh,t-1}$ and 0 otherwise

5 x_{ht} is total household weekly expenditure on foods per household member

6 P_{ht} is a Laspeyres price index u_{iht} is the error term

7 This empirical specification is obtained from a PIGLOG cost function incorporating
 8 reference prices, based on Putler (1992), where the reference prices are equal to the prices
 9 faced in the previous shopping trip (Putler, 1992). More intuitively, the price effect in (1) is
 10 the combination of three dimensions: the average price effect is captured by the parameter
 11 γ_{ij} , the ‘loss’ dimension is embodied by the parameter δ_{ij} and the ‘gain’ dimension by the
 12 parameter ω_{ij} . Depending on whether the individual consumer in a given time period is
 13 experiencing a loss ($p_{jht} > p_{jh,t-1}$) or a gain ($p_{jht} < p_{jh,t-1}$), the resulting price coefficient
 14 that will feed into the elasticity equation will be $\gamma_{ij} + \delta_{ij}$ or $\gamma_{ij} + \omega_{ij}$, respectively. The sign
 15 before ω_{ij} is reverted relative to the specification in (1) because when a gain is incurred we
 16 have that $p_{jht} - p_{jh,t-1} < 0$.

17 To reduce possible endogeneity between expenditure share (w_{iht}) and total expenditure
 18 (x_{ht}) we regress household per capita food expenditure on household socio-demographic
 19 characteristics (tenure, income, income squared, household size, presence of children) and 4-
 20 weekly dummies for seasonality, and use predicted values from the regression as instruments
 21 for total expenditure (Blundell et al., 1993).

22 Our base unit is household-week and as we model a large number of relatively disaggregated
 23 food groups, we observe a significant proportion of non-purchases as not all households

1 purchase items in each food group on weekly basis. To correct for the potential bias which
 2 may arise from including the zero-purchases we adopt a two-step approach by Shonkwiler
 3 and Yen (1999) (Shonkwiler & Yen, 1999) which is regularly used in food demand studies
 4 (Caro et al., 2017; Ecker & Qaim, 2011). In the first step, the decision to purchase foods in
 5 any food group in a given week by a household is modelled using a probit model. The
 6 decision (a binary variable equal to one if expenditure is positive) is modelled as a function of
 7 household socio-demographic variables: household size, presence of children, age of the
 8 main respondent, socio-economic status, tenure, income and education level of the main
 9 respondent, lagged volume of purchases to account for possible stockpiling, and monthly
 10 time dummies to take into account seasonal effects. From the probit model, we estimate the
 11 probability density function (ϕ_i) and cumulative density function (Φ_i) of the linear
 12 predictions of the fitted model. These are applied in the second step (2) to estimate the
 13 demand function in (1) which accounts for the probabilities of purchase, on the full panel-
 14 data set:

$$15 \quad w_{iht}^* = \Phi_{iht}(w_{iht}) + \varphi_i \phi_{iht} + \sum_{t=1}^{26} \rho_{it} T_{it} + v_{ih} + \varepsilon_{it} \quad (2)$$

16 Where T_{it} are dummy variables, added to capture any seasonal or other time effects (n=26 4-
 17 week periods) and v_{ih} is a fixed household effect. These fixed effects act as shifter of the
 18 “average” demand model intercept α_i and the model is estimated with a fixed effects
 19 estimator.

20 For each food group $i=1, 2, \dots, 26$ we estimate (2) with robust standard errors to allow for any
 21 misspecification, particularly serial correlation of observations within the households.
 22 Standard errors are further adjusted for clustering to the geographical area used in estimating
 23 prices (n=110).

1 Marshallian demand elasticities (% response in demand to 1% change in price) unconditional
 2 of purchase, were calculated for price increases (e_{ij}^{UP}) and decreases (e_{ij}^{DOWN}) (Yen et al.,
 3 2002):

$$4 \quad e_{ij}^{UP} = \Phi_i * \left(\frac{\gamma_{ij} + \delta_{ij}}{w_i} - \frac{\beta_i w_j}{w_i} \right) - \Delta_{ij}$$

$$5 \quad e_{ij}^{DOWN} = \Phi_i * \left(\frac{\gamma_{ij} - \omega_{ij}}{w_i} - \frac{\beta_i w_j}{w_i} \right) - \Delta_{ij}$$

6

7 **2.2 Estimation of prices**

8 The dataset includes ~75m individual item purchases by households over a two-year period.
 9 For manageable estimation we divide products into 26 categories of foods and beverages and
 10 aggregate individual item purchases to weekly expenditures and quantities in each group
 11 from which we calculate weekly unit prices as a ratio of the two. Such unit values are known
 12 to suffer from quality and aggregation bias and should not be used directly (Deaton, 1988).
 13 To address this we assume that in a relatively small geographical area households face the
 14 same prices in the same time period (Deaton, 1988) and estimate average weekly unit prices
 15 for each of the 26 food and beverage groups in postcode areas (n=110) observed in the data.
 16 A specific feature in the UK is that food retailers and the UK Competition Commission have
 17 an agreement on national pricing policy (Griffith et al., 2010; UK Competition Commission,
 18 2008). The agreement implies that large chain supermarkets apply same pricing (and
 19 promotions) in same types of branches across its chain nationally (i.e. price may vary
 20 between different types of branches but not within one type of branches). Price variation in
 21 food groups by geographical area could occur across the regions due to differences in the
 22 distribution of shop types or product range in the shops, of which some could be due to local
 23 demand shocks. However, given our time unit is one week, we can safely assume that such

1 demand shocks do not influence the concentration of shop types in the area or the product
2 range available within this short time unit. Thus, the geographical average weekly prices we
3 use are assumed to be exogenous to local demand.

4 Household-weeks where no expenditure on any foods or beverages were made were excluded
5 and thus the difference in the current and previous purchase price for any single household is
6 based on the difference in the price between purchase occasions rather than calendar weeks.
7 Finally, to avoid introducing marginal price changes through the process of estimating
8 average geographical prices, we limit price increases or decreases to only changes with
9 difference between current and previous purchase price greater than 5%.

10 **2.3 Data**

11 The household expenditure data comes from a large commercial home-scan panel (~32,000
12 households annually) operated by Kantar Worldpanel. Our dataset covers the period from
13 January 2012 to December 2013. The panel is nationally representative of the population in
14 Great Britain with respect to household size, number of children, social class, geographical
15 region and age group. Participants to the panel are recruited by Kantar Worldpanel via postal
16 mail and e-mail and panel representativeness is assessed by Kantar Worldpanel at 4-week
17 intervals. Households supply data on items purchased and brought home by scanning
18 barcodes of the bought products and by sending in digital images of till receipts. Households
19 are additionally supplied with barcodes to record purchases of unpackaged products, such as
20 fruits and vegetables. Our dataset covers information on purchases (volume and expenditure)
21 of food and drinks from large retailers, supermarkets, butchers, greengrocers, and corner
22 shops. The data exclude all purchases of food and beverages not brought home (e.g.
23 consumed out of homes). Participants are offered vouchers for retailers and for leisure
24 activities of an average value of less than £100 per household per year.

1 Socio-demographic data are collected annually and describe household size and presence of
2 children, age, and highest qualification of the main shopper, geographical location (postcode
3 area), income group, occupational socio-economic status (SES) and tenure of the household.
4 For the analysis we divide households by SES into three groups: high-SES (A&B: higher
5 managerial and professional workers), mid-SES (C1&C2: white collar and skilled manual
6 workers), and low-SES (D&E: semi-skilled, unskilled and retired worker) households.

7 The data also include information on nutritional composition of each of the foods purchased:
8 energy (kcal), carbohydrates (g), sugar (g), fats (g), sodium (g), fibre (g), and protein (g).

9 These data are compiled by Kantar Worldpanel based on information on product packaging.

10 Although Kantar Worldpanel does not collect fruit and vegetable content of products, they
11 have provided a score (low – 0, some – 1 and high– 5) for content of fruit and vegetables.

12 **2.4 Food groups**

13 Food groups are first created following the market classifications in the dataset. In order to
14 classify foods and beverages into less healthy and healthier products we use a nutrient
15 profiling model (see supplementary material [INSERT LINK TO ONLINE FILE A]) (UK Department
16 of Health, 2011). The underlying principle is that energy, sugar, sodium and saturated fats
17 contribute to ‘positive’ points and fibre, protein and fruit and vegetable content contribute to
18 ‘negative’ points, which are added together. The higher the score, the less healthy the product
19 is. We apply the cut-off point suggested in the guidance and consider a product to be less
20 healthy if it scores above four points or a drink scores above one (UK Department of Health,
21 2011).

22 Based on the profiling model, eight food groups (bread & morning goods; cereal & cereal
23 bars; dairy; fresh & frozen red meat; processed meat & fish; ready meals & convenience
24 foods; sauces & condiments; non-alcoholic beverages) were divided into healthier and less

1 healthy categories (see table 1 in supplementary material for detailed description [INSERT LINK
 2 TO ONLINE FILE A]). For the remaining healthier (pasta, rice, grains, dry pulses; eggs, fresh &
 3 frozen fish & white meat; fruits; vegetables) and less healthy food groups (fat & oil; savoury
 4 snacks; sweet snacks; desserts & puddings; alcohol and other foods (including table salt and
 5 sugar)) such differentiation was not made as most products in each (>80%) were either
 6 healthier or less healthy.

7 **2.5 Simulations of changes in nutrient purchases**

8 Assuming a linear price response, we apply an increase of 20% on the price of all less healthy
 9 food groups (where explicitly specified) and on fat & oil, savoury snacks, sweet snacks,
 10 desserts & puddings, other foods and alcohol. A decrease of 20% is applied on the price of
 11 healthier food groups (where specified); pasta, rice, grains & dry pulses; fresh & frozen white
 12 meat & fish, eggs; fruits; and vegetables. The outcomes measured are changes in the energy
 13 (kcal) and nutrient (sugar (g), saturated fats (g), sodium (g), protein (g) and fibre (g)) content
 14 of average per capita daily take-home purchases, due to price changes in individual food
 15 groups and in all groups combined.

16 For each outcome measure k we estimate the change in purchases Z_k due to the change in the
 17 price in food group i :

$$Z_k = \sum_{i,j=1}^{26} x_{ik} * e_{ij}^{up/down} * \Delta p_i$$

18 Where e_{ij} is price elasticity (increasing or decreasing price) and Δp_i is the price change of
 19 20% (either increase or decrease). Baseline purchases of energy or each nutrient (per
 20 capita/day) in each food group (x_{ik}) in 2013 were estimated by first aggregating individual
 21 purchases in 2013 dataset to weighted population purchases which we then divide by number
 22 of days in the year (365) and population size. The latter is calculated from data provided by

1 Kantar Worldpanel on the number of households and household size (including by SES) in
2 Great Britain. Kantar Worldpanel provided gross-up weights that account for sampling and
3 non-response in the panel.

4

5 **3. Results**

6 **3.1 Household demographics and food purchases**

7 The final sample includes 2,057,204 weekly purchase observations from 26,799 households.
8 Due to missing values for income, we had to exclude 392,818 weekly observations. Even
9 with this deletion, the distribution of households over SES is broadly in line with national
10 figures (see table 1 for descriptive statistics). The number of weeks households had positive
11 expenditure ranged from 1 to 104, with on average 72 weeks over the two years.

12 *Table 1 here*

13 Table 2 describes the share of non-zero observations, weekly expenditure shares and average
14 prices. On average, 53% of expenditures was towards healthier foods. Across the food
15 groups, households spent most on vegetables (9%, average weekly expenditure £4.70),
16 healthier ready meals and convenience foods (8%, £4.10), alcohol (7%, £5.01), sweet snacks
17 (7%, £3.30) and healthier dairy (7%, £2.90).

18 *Table 2 here*

19 Across the SES groups (table 2 in supplementary material) [INSERT LINK TO ONLINE FILE A],
20 high-SES households spent relatively more of take-home food expenditure on healthier foods
21 (55% in comparison to 52% for low-SES), including more on fruits and vegetables, and fresh
22 & frozen white meat, fish, eggs. Low-SES households spent overall relatively more on less
23 healthy foods (48% in comparison to 45% for high-SES), including on sweet snacks but they

1 also spent relatively more on healthier bread & morning goods and healthier ready meals &
2 convenience foods.

3 Highest average price (per Kg/L) was observed for savoury snacks (£6.63), fresh and frozen
4 red meat (£6.34, £6.15), and less healthy ready meals and convenience foods (£5.88).
5 Healthier dairy (£0.84), less healthy beverages (£1.12), healthier bread & morning goods
6 (£1.17), healthier beverages (£1.33) and vegetables (£1.59) had the lowest average unit
7 prices. Average price difference between current and previous price was £0.11 with average
8 increase and decrease only marginally different. Largest price differences between purchases
9 occasions were among healthier sauces & condiments (£0.50-0.57), fresh & frozen white
10 meat, fish eggs (£0.10-0.11), less healthy bread & morning goods (£0.10), and healthier
11 drinks (£0.07).

12 **3.2. Nutrient composition of purchases**

13 A summary of average energy and nutrient content of daily per capita purchases is shown in
14 table 3, and by food groups in table 4. On average, households purchased 2,111 kcal of
15 energy per day per capita. Low-SES household purchases yielded to slightly more energy
16 (2,161 kcal) in comparison to mid- and high-SES households (2,078 kcal and 2,119 kcal per
17 capita/day, respectively).

18 *Table 3 here*

19 When comparing purchases to reference daily intake level (RDI) across the SES, purchases of
20 sugar, sodium, saturated fat and protein were well above the RDI while purchases of fibre
21 were close to reference level. On average, mid-SES households were closest to the RDI level.

22 Across the food groups, sorted by the average nutrient profile score (table 4 and tables 3-5 by
23 SES in supplementary material [INSERT LINK TO ONLINE FILE A]), purchases contributing most

1 to energy were healthier bread & morning goods (229.1 kcal), desserts & puddings (205.5
2 kcal), sweet snacks (202.1 kcal) and fat & oil (159.8 kcal).

3 *Table 4 here*

4 Fat & oils were the biggest source of saturated fats followed by sweet snacks, desserts &
5 puddings and less healthy dairy. Desserts & puddings, and sweet snacks were the largest
6 sources of sugar purchases, followed by other foods (includes table sugar), healthier dairy
7 and fruits. Sodium purchases were largest in other foods (includes table salt), healthier bread
8 & morning goods, less healthy processed meat & fish, and in ready meals & convenience
9 foods.

10 **3.3. Demand elasticities and asymmetry**

11 Figure 1 presents the own-price elasticities (food groups on the x-axis are sorted based on the
12 average nutrient profile score starting from healthiest on the left hand side on both panels)
13 with a linear trend through the estimates. Overall, the own-price elasticities range from -0.380
14 to -1.074 (the full set of price elasticities with confidence intervals, are in supplementary
15 material [INSERT LINK TO ONLINE FILE A] tables 6-9). On average, household response is
16 stronger to price increases, with mean own-price elasticity for increasing prices of -0.847 in
17 comparison to that for decreasing prices of -0.780. From the 26 food groups, ten have a
18 statistically significant difference ($p < 0.05$) between the elasticity of price increases and
19 decreases with an average difference of 0.119. This implies that if price increases by 10%
20 there would be a 1.19 percentage point greater response in demand compared to a price
21 decrease of 10% for the same food. The largest asymmetry (0.313) was in the demand for less
22 healthy fresh & frozen red meat. If comparing based on healthiness of the foods, own-price
23 elasticities for healthier foods tended to have on average a smaller asymmetry (0.047) in
24 comparison to less healthy foods (0.084).

1 High-SES households were less sensitive to price changes in comparison to mid-SES and
2 low-SES households. Low-SES households also had greater asymmetry in own-price
3 elasticity (on average 0.183) in comparison to high-SES (0.074).

4 *Figure 1 here*

5 Looking across the x-axis in both panels on Figure 1 where foods on the left are relatively
6 healthier and less healthy on the right, it can be seen that the gap between the price response
7 of high-SES and low-SES households becomes wider. For increasing prices, this gap is
8 driven more by low-SES group being more price responsive as foods become less healthy
9 (high-SES response stays relatively stable), whereas for decreasing prices the gap is mainly
10 driven by high-SES group becoming relatively less responsive as foods become less healthy
11 (low-SES household price response to decreasing prices stays relatively stable).

12 *Cross-price effects*

13 The magnitudes of cross-price elasticities were generally small, with most within $|0.2|$ with a
14 mix of (complement) income and substitution effects (see supplementary material [INSERT
15 LINK TO ONLINE FILE A] tables 6-9). Price changes of sweet snacks, and healthier dairy affected
16 the demand for largest number of groups (20 and 19 significant cross-price effects, $p < 0.05$,
17 respectively). Price changes in less healthy fresh and frozen red meats, and in fat and oil
18 affected the least (3-4 significant cross-price effects, $p < 0.05$).

19 Of the 650 pairs of cross-price elasticities, on average 36-37% were significant at 5% level
20 (242 for increasing and 234 for decreasing prices). Of these 45 (18%) had a statistically
21 significant difference in the elasticity of price increase and decrease ($p < 0.05$). The majority
22 of the asymmetries were due to only one of the elasticities (of either price increase or
23 decrease) being statistically different from zero. The average magnitude of the asymmetry

1 was -0.1 where it was negative and 0.08 where positive. Considering the overall magnitudes
2 of cross-price effects being generally small, these are relatively large differences.

3 **3.4 Simulation of price changes**

4 Figure 2 shows how energy and nutrient content of purchase change following a 20% price
5 increase in each of the less healthy food groups and a 20% price decrease in each of the
6 healthier food groups individually. Full results, including by SES with confidence intervals
7 are shown in table 10 in the supplementary material [INSERT LINK TO ONLINE FILE A].

8 Combining own- and cross-price effects, we find the largest reduction in energy purchases
9 per capita/day when the price of sweet snacks (46.0kcal 95%CI -55.3 to -36.7kcal), desserts
10 & puddings (38.8kcal 95%CI -47.3 to -30.4kcal), and fat & oil (28.5kcal 95%CI -37.1 to -
11 20.0kcal) increases. Purchases of sugar decreased most if the price of desserts & puddings,
12 sweet snacks, and 'other' foods (including table sugar) increases: -4.0g (95%CI -4.9 to -3.2g),
13 -3.8g (95%CI -4.6 to -2.9g), and -2.4g (95%CI -3.0 to -1.8g), respectively. Reduction in
14 saturated fat purchases was greatest if price of fat & oil increased (-1.0g 95%CI -1.2 to -0.7),
15 followed again by sweet snacks (-0.9g 95%CI -1.1 to -0.6g) and dessert & puddings (-0.7g
16 95%CI -0.8 to -0.5g). In all of the above food groups, the reduction in purchases is relatively
17 larger among low-SES households. Sodium purchases reduced most through increased price
18 of 'other' foods (including table salt) (-0.08g 95%CI -0.11 to -0.06g), processed meat & fish
19 (-0.05g 95%CI -0.06 to -0.03g), less healthy ready meals & convenience foods as well as less
20 healthy sauces and condiments (for both -0.04g 95%CI -0.05 to -0.03g).

21 In terms of trade-offs with fibre and protein, price increases reduce purchases of fibre most
22 from sweet snacks and desserts & puddings (0.33g 95%CI -0.37 to -0.28g and -0.27g 95%CI
23 -0.32 to -0.22g, respectively). Protein purchases are reduced most by increasing the price of

1 less healthy dairy (1.1g 95%CI -1.3 to -0.9g) and less healthy processed meat & fish or less
2 healthy ready meals and convenience foods (1.0g 95%CI -1.2 to -0.7g in each).

3 Greatest changes in the purchases of fibre were observed if the price of vegetables and
4 healthier bread & morning goods decreases (by 0.6g in each (95%CI 0.4 to 0.8g)). However,
5 a reduction in prices of healthier foods leads also to higher energy and purchases of other
6 nutrients. The largest effect in energy purchases is observed when the price of healthier bread
7 & morning foods decreases (energy purchases increase by 34.8kcal 95%CI 23.7-45.9kcal). A
8 price decrease in this group would also contribute, relative to other food groups, to a larger
9 increase in the purchases of sodium (0.05g 95%CI 0.03-0.07g). Sugar purchases would also
10 increase, particularly from the decrease of price of fruits (1.7g 95%CI 1.1-2.4g) and among
11 high-SES households who are relatively more responsive to changes in the price of fruits
12 (2.5g 95%CI 1.4-3.6g). However, reduction in the price of fruit would have an additional
13 benefit, through cross-price effects, of reducing purchases of saturated fats (-0.13g 95%CI -
14 0.06 to -0.2g), while for vegetables a price decrease would lead to greater purchases of
15 saturated fats through substitution effects (0.47g 95%CI 0.11-0.55g). To understand the
16 impact on purchases of vegetables and fruit in daily portions, we can use the estimated
17 increase in calories from own-price effects (18kcal and 9kcal, respectively). Assuming a
18 portion of 80g of fruits or vegetables has approximately 50-60 kcal, the 20% price reduction
19 in both would increase purchases by approximately half a portion per day/person.

20 Finally, table 5 shows the extent of changes in purchases of energy and nutrients if prices of
21 all less healthy foods as defined above increased, and prices of all healthier foods decreased.

22 *Table 5 here*

23 In the full sample such a scenario would lead to a net reduction of 67.6kcal purchased per day
24 per capita which increases to 91.3kcal in low-SES households reflecting larger asymmetry in

1 price response in that group as well as greater responsiveness to price increases in less
2 healthy foods. Alongside energy, the sugar, saturated fat and sodium content would also
3 decrease whereas protein and fibre content would increase. With respect to household
4 expenditure, such price changes led to a small reduction in the total expenditure (£1.37 per
5 day) with smallest saving (£0.95) for low-SES and greatest saving for mid-SES households
6 (£0.145).

7 **4. Robustness analysis**

8 Given the modifications done in the AIDS modelling approach, we carried out two analyses
9 of robustness of the simulation results. First, we applied elasticities that are estimated from a
10 model applying two restrictions - adding-up and homogeneity on the parameters in the
11 demand model in (1) which ensure theoretical consistency of the AIDS model. The panel
12 structure of the data, the consideration of fixed effects, and the large number of food groups
13 in our model make it unfeasible to test or impose cross-price symmetry across equations.
14 While this is certainly a limitation, the generalization of the demand model, allowing
15 different response to cross-price increases and decreases, also makes the requirement of
16 symmetric cross-price coefficients less straightforward. Using elasticities from the model
17 with restrictions, there were marginal differences in the estimated changes in energy and
18 nutrient purchases (see table 11 in supplementary material [INSERT LINK TO ONLINE FILE A]).
19 Assessing the confidence intervals, these differences were significant for protein purchases
20 from four food groups and sugar purchases from two food groups.

21 Second, we applied elasticities estimated from a symmetric demand model in the simulation.
22 Using these would be expected to lead to smaller differences between price increases and
23 decreases. Again, the differences in simulation results are marginal, with significant
24 differences only in a handful of food groups (sugar purchases in five food groups, saturated

1 fat and protein purchases in two groups and fibre in three food groups). Summing the
2 differences across all food groups, as expected, the symmetric model estimated higher effect
3 for price decreases and lower effect for price increases.

4

5 **5. Discussion**

6 This paper contributes to the literature by exploring the effects of price increases and
7 decreases on purchases of energy and nutrients from a large number of healthier and less
8 healthy food groups. While taxes on SSBs in particular have received recent policy focus, it
9 is important to continue considering the whole diet, including how to increase consumption
10 of healthier foods.

11 We found, that on average, the energy and nutrient content of take-home food and beverage
12 purchases were generally above the reference daily intake level, and this is without
13 considering any foods or beverages consumed out-of-homes, which can count up to 28% of
14 total food and beverage expenditures (The Food Foundation, 2018). Low-SES spent, on
15 average, a greater share of expenditures on less healthy foods (48%) in comparison to high-
16 SES (45%), who in turn spent relatively more on healthier foods.

17 Based on simulations by individual food groups, a price increase of 20% would reduce
18 purchases of energy most among sweet snacks or desserts & puddings (46kcal and 39kcal per
19 capita/day, respectively). In addition, sugar purchases would drop by 3.7g and 4.0g per
20 capita/day respectively. This compares to the average of 3.2kcal and 1.1g sugar reduced from
21 increasing the price of less healthy beverages that are subject to recent policies in Britain and
22 other countries. However, increasing the price of sweet snacks and desserts also lead to a
23 reduction in fibre and protein purchases to a greater extent, in comparison to less healthy

1 beverages. An additional 2.2g of sugar could be reduced by increasing the price of caloric
2 sweeteners (table sugar, honey, syrup).

3 Sodium purchases were reduced most through increases in the price of table salt (0.8g), but
4 even higher reductions (0.12g) could be achieved by increasing the price of less healthy
5 processed meat, ready meals, and table sauces. All these changes were found to be higher
6 among low-SES households, driven from greater price sensitivity when foods are less
7 healthy, and also a slightly higher baseline contribution of less healthy foods.

8 The trade-offs were apparent when reducing the price of healthier foods. For example
9 increasing fibre purchases via reduced price of healthier bread and morning foods would also
10 lead to the greatest increase in energy and sodium purchases. Decreasing the price of
11 vegetables led to the greatest increase in saturated fats purchases (through cross-price
12 effects). Purchases of fibre from fruits would increase relatively less if its price reduced but
13 this would also lead to substitution effects towards less saturated fats purchased. These
14 findings indicate a complex picture on how changes in prices of different foods may affect
15 food and nutrient purchases and the need to consider the full range of cross-price effects.

16 Combining price increases on all less healthy food groups with a price decrease on all healthy
17 groups showed that diet overall is likely to improve from such changes, with a reduction in
18 energy purchases by 55-91kcal, and sugar purchases by 5-9g, with the higher values seen for
19 low-SES group. Fibre purchases would increase by 0.8-1.1g (4-5% of RDI), with the largest
20 effect in the mid-SES group, as well as protein purchases by 3-4g. Protein purchases however
21 are already well above (by 20g on average) the RDI level (50g), which suggest that this
22 change is not necessarily positive. Combined price changes would lead to an average of £1.37
23 saved per day per household, with the greatest savings observed among mid-SES households.

1 When looking at how much households spend on average, and how frequently, it is striking
2 that sweet snacks (i.e. biscuits, confectionary, chocolate) have a relatively high expenditure
3 share (7%) and households buy these frequently (69% of week) even though its shelf-life is
4 relatively long. As a comparison, 6% was spent on fruits, which were bought in 67% of the
5 weeks. High-SES households purchased more fruits with higher expenditure share (7% vs
6 5%) and energy purchased (76kcal vs 58kcal) per capita/day in comparison to low-SES
7 households. To the contrary, low-SES households spent relatively more on sweet snacks (8%
8 vs 6%) and purchased more energy from these (226kcal vs 182kcal). These findings,
9 however, do not mean that low-SES households necessarily have a worse diet, as the data
10 exclude purchases for consumption outside of homes, which are likely to be higher among
11 mid- and high SES households due to higher average earnings(The Food Foundation, 2018).

12 Our estimates are consistent with existing literature modelling the demand for foods and
13 beverages in the UK (Green et al., 2013). For example based on data from 2009, own-price
14 elasticity for dairy & egg was reported -0.505, meat -0.804, fish -0.441, fruits & nuts -0.698,
15 vegetables -0.633, and fats & starches -0.847 (Tiffin et al., 2011). In comparison to
16 simulation studies of price changes we find the same results as Briggs et al. (2013) who
17 modelled a 3kcal reduction per capita/day reduction from a 20% tax on sugar-sweetened
18 beverages (Briggs et al., 2013). Another study using expenditure data from 2002-2006
19 applied as one of the strategies a VAT style tax of 17.5% on less healthy foods based on a
20 'WXYfm' nutrient profile index, finding a reduction in energy intake on average by 2.4% and
21 of saturated fats by 3.1% (Nnoaham et al., 2009). When we sum the effect of price increases
22 across the less healthy foods, we see an 11% decrease in daily energy and 13% in saturated
23 fats purchases, which considering different data, years under study and greater price change,
24 are relatively consistent.

1 With respect to asymmetry, the results agree with marketing research that consumer response
2 is generally stronger to price increases (Hardie et al., 1993; Koszegi & Rabin, 2006; Pauwels
3 et al., 2007; Putler, 1992). For individual food groups the difference between the impact of
4 price change when using symmetric demand model for elasticities in comparison the
5 asymmetric model, was small and generally towards symmetric model providing a greater
6 effect for price decreases and smaller effect for price increases. This result is intuitive as it
7 can be thought of providing an average response over increasing and decreasing prices. Such
8 asymmetry has implications for policy suggesting that subsidies need to be relatively larger in
9 magnitude in comparison to taxes to achieve an equivalent change in demand.

10 Our analysis has a number of strengths. First, we were able to analyse the demand through
11 food groups but apply simulations directly to nutrient purchases, which is more relevant when
12 considering potential health impacts of changing food prices and thus fiscal policy. Second,
13 we take into account household fixed effects controlling for heterogeneous and diverse
14 consumption patterns for each specific food. By incorporating price differences across
15 purchase occasions we allow for greater flexibility in consumers response to decreasing and
16 increasing prices. The nutrient profile model we use has been used in existing policies
17 (Mayor of London, 2018; Office of Communications, 2007) and continues to be used in
18 health-related food policies put forward (Department of Health & Social Care, 2019),
19 strengthening its applicability in real settings.

20 Some limitations to the analysis need to be considered. We do not impose restrictions
21 assumed in the traditional AIDS framework and estimate the model equation-by-equation,
22 rather than as a system. We do this to be able to include fixed-effects in a relatively simple
23 framework, as household heterogeneity in the marketing literature has been argued as a
24 potential driver of asymmetry and we considered this crucial to control for. We tested the
25 robustness of simulation results by using elasticity estimates from a model with adding-up

1 and homogeneity restrictions applied and found only small changes in the simulation
2 estimates. The impact of not imposing the symmetry restriction on cross-price elasticities is
3 more difficult to disentangle, as we cannot run a robustness model with this restriction
4 imposed without formidable estimation difficulties and thus this should be considered in the
5 interpretation of cross-price elasticities. To the best of our knowledge, all previous studies
6 allowing for a different response to price increases and decreases were estimated on an
7 equation-by-equation basis, and without imposing symmetry on cross-price coefficients.
8 Given our adjustment to the standard modelling approach to allow for asymmetric price
9 response, it remains an open research question whether and how symmetry should be tested
10 and imposed. We also estimate separately, rather than jointly, the first step probit equations to
11 address biases related to zero expenditures. This does not affect the consistency of our
12 estimated parameters, and makes our estimates less efficient, but given the large sample we
13 do not believe this limitation to question our findings.

14 We excluded ~15% of observations due to missing household income value. Regardless, the
15 sample distribution across SES remains relatively similar to the distribution of households
16 across SES in the population. We did not have information on the age and sex of all
17 household members and therefore could not adjust purchases to household composition.
18 Finally, as we do not have food waste estimates our results are more likely to be biased
19 upwards for actual consumption.

20 In conclusion, the analysis demonstrates that on average energy and nutrient content of take-
21 home food purchases is already higher than recommended for total daily intake, without
22 considering food consumed out-of-homes. However, this can be improved, without
23 detrimental consequences on food expenditures, if the price of less healthy foods increased
24 and the price of healthy foods decreased, with relatively bigger improvements seen in the
25 nutritional quality of purchases of low-SES households.

1

2

3

ACCEPTED MANUSCRIPT

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25

Table 1. Household demographics

	Low-SES	Mid-SES	High-SES	Full sample
Population (2013 in Great Britain)	17,219,584 (27%)	30,508,779 (49%)	14,666,726 (24%)	62,395,091
Number of households in sample (%)	6,200 (23%)	14,938 (56%)	5,661 (21%)	26,799
	Average			
Age of main shopper (SD)	51.4 (15.1)	48.6 (14.8)	48.7 (15.0)	49.3 (15.0)
Household size (SD)	2.5 (1.4)	2.7 (1.3)	2.8 (1.3)	2.7 (1.3)
Share (%) of households with children (SD)	0.3 (0.5)	0.4 (0.5)	0.4 (0.5)	0.4 (0.5)
Average number of children (if have) (SD)	1.9 (0.9)	1.8 (0.8)	1.8 (0.8)	1.8 (0.8)
Single person households (%)	1,608 (26%)	2,640 (18%)	926 (16%)	5,177
Income	% by SES			
up to 20,000 pa	71.3	30.12	10.3	
20-49,000 pa	27.6	57.3	51.1	
> 50,000 pa	1.3	12.5	38.6	
Highest qualification				
Degree or higher	9.3	25.1	60.7	
Higher education	12.9	18.5	13.6	
A Level	11.5	16.5	10.1	
GCSE	31.3	24.0	9.6	
Other	13.2	9.3	4.2	
None	21.8	6.6	1.8	
Tenure				
Owned outright	22.4	27.2	31.7	
Mortgaged	21.8	45.9	54.4	
Rented	54.2	25.4	12.8	
Other	1.6	1.5	1.1	
Region				
London	12.3	15.5	18.1	
Midlands	15.1	14.4	15.9	
North East	5.7	4.8	4.7	
Yorkshire	14.6	13.8	12.7	
Lancashire	11.2	10.5	10.0	
South	10.4	10.8	11.5	
Scotland	9.3	8.9	7.9	
East of England	9.0	9.0	8.2	
Wales and West England	8.8	8.6	8.3	
South West	3.6	3.7	2.7	

Table 2. Expenditure shares, prices and extent of censoring in the data

n=2,057,204	Share of non-zero observations		Expenditure share		Price (£) ¹	
	Mean	SD	Mean	SD	Mean	SD
Bread & morning goods (healthier)	0.77	0.42	0.05	0.07	1.17	1.06
Bread & morning goods (less healthy)	0.31	0.46	0.01	0.03	1.86	1.15
Cereals & cereal bars (healthier)	0.22	0.41	0.01	0.03	2.99	1.11
Cereals & cereal bars (less healthy)	0.28	0.45	0.02	0.04	4.26	1.08
Pasta, rice, grains, dry pulses	0.34	0.48	0.01	0.03	1.93	1.12
Dairy (healthier)	0.80	0.40	0.07	0.09	0.84	1.07
Dairy (less healthy)	0.59	0.49	0.04	0.06	4.93	1.07
Fresh & frozen white meat & fish, eggs	0.56	0.50	0.06	0.08	1.39	1.19
Fresh & frozen red meat (healthier)	0.22	0.42	0.02	0.06	6.34	1.09
Fresh & frozen red meat (less healthy)	0.16	0.37	0.01	0.04	6.15	1.10
Processed meat & fish (healthier)	0.29	0.45	0.02	0.04	5.09	1.10
Processed meat & fish (less healthy)	0.43	0.50	0.03	0.06	4.90	1.08
Ready meals & convenience foods (healthier)	0.71	0.45	0.08	0.10	3.32	1.08
Ready meals & convenience foods (less healthy)	0.66	0.47	0.05	0.07	5.88	1.07
Fruits	0.67	0.47	0.06	0.08	1.75	1.09
Vegetables	0.83	0.37	0.09	0.09	1.59	1.10
Sauces & condiments (healthier)	0.19	0.39	0.01	0.02	4.98	1.28
Sauces & condiments (less healthy)	0.35	0.48	0.01	0.03	3.63	1.11
Fat & oil	0.41	0.49	0.02	0.04	2.91	1.08
Non-alcoholic drinks (excl. dairy) (healthier)	0.64	0.48	0.06	0.08	1.33	1.16
Non-alcoholic drinks (excl. dairy) (less healthy)	0.33	0.47	0.02	0.05	1.12	1.12
Savoury snacks	0.49	0.50	0.03	0.05	6.63	1.05
Sweet snacks	0.69	0.46	0.07	0.10	1.40	1.17
Desserts & puddings	0.70	0.46	0.06	0.08	2.09	1.09
Other foods	0.24	0.43	0.01	0.04	1.75	1.14
Alcohol	0.30	0.46	0.07	0.15	4.68	1.15
Total/average	0.47		£52.4		£3.27	

Notes: ¹Average geographical unit value

Table 3. Energy and nutrient content of daily per capita take-home purchases

Energy/nutrient (reference daily intake)	Full sample	Low-SES (D&E)	Mid-SES (C1&C2)	High-SES A&B
Energy (2,000kcal)	2,111	2,161	2,078	2,119
Sugars (90g)	117.5	120.9	114.7	119.5
Sodium (<2.3g; eqv. <6g of salt)	2.8	2.9	2.7	2.7
Saturated fat (20g)	32.1	33.1	31.5	32.1
Fibre (18g)*	18.2	17.7	17.9	19.3
Protein (50g)	70.2	70.5	69.3	71.6

* non-starch polysaccharides (NSP) fibre

Table 4. Energy and nutrient content of daily per capita take-home purchases by food group

	NPM score¹	Energy (kcal)	Sugar (g)	Sod.² (g)	Sat. fat (g)	Fibre³ (g)	Prot. (g)
Vegetables	-7.3	116.1	5.51	0.07	0.37	4.24	4.56
Fruits	-4.2	64.1	13.12	0.00	0.06	1.51	0.89
Cereals & cereal bars (healthier)	-2.9	40.4	1.10	0.01	0.10	1.00	1.14
Bread & morning goods (healthier)	-1.6	229.1	3.63	0.34	0.80	3.55	8.57
Non-alc. drinks (healthier)	-1.3	26.7	5.50	0.01	0.05	0.27	0.34
Pasta, rice, grains, dry pulses	-1.1	67.4	0.38	0.03	0.14	0.67	2.08
Fresh & frozen red meat (healthier)	-0.4	24.2	0.01	0.01	0.56	0.02	3.24
Ready meals & conv. foods (healthier)	-0.4	105.9	1.85	0.18	1.11	1.42	4.36
Sauces & condiments (healthier)	-0.3	3.7	0.24	0.02	0.03	0.08	0.09
Dairy (healthier)	0.1	87.2	0.12	0.09	1.32	0.11	10.04
Fresh & frozen white meat & fish, eggs	0.1	141.1	13.83	0.12	3.17	0.27	9.20
Processed meat & fish (healthier)	0.4	20.1	0.17	0.04	0.21	0.12	1.94
Alcohol	0.4	73.9	1.93	0.01	0.00	0.00	0.20
Non-alc. drinks (less healthy)	3.3	27.6	5.87	0.03	0.07	0.05	0.11
Desserts & puddings	9.5	205.5	21.23	0.08	3.73	1.21	2.86
Fresh & frozen red meat (less healthy)	9.6	22.3	0.04	0.02	0.72	0.01	1.92
Cereals & cereal bars (less healthy)	10.1	49.2	3.26	0.04	0.29	0.68	0.93
Other foods (incl. table salt and table sugar) ⁴	10.2	63.3	14.51	0.57	0.02	0.15	0.31
Savoury snacks	10.7	76.4	0.63	0.10	0.67	0.64	1.34
Bread & morning goods (less healthy)	10.8	51.2	1.09	0.06	0.94	0.35	1.21
Ready meals & conv. foods (less healthy)	10.9	70.4	1.05	0.19	1.54	0.48	3.13
Sauces & condiments (less healthy)	12.8	29.0	1.98	0.18	0.28	0.11	0.28
Proc. meat & fish (less healthy)	14.7	64.8	0.29	0.24	1.69	0.21	4.54
Dairy (less healthy)	16.2	89.3	1.95	0.13	4.45	0.08	4.36
Sweet snacks	19.7	202.1	18.11	0.08	4.60	0.93	2.46
Fat & oil	21.0	159.8	0.11	0.10	5.20	0.03	0.09

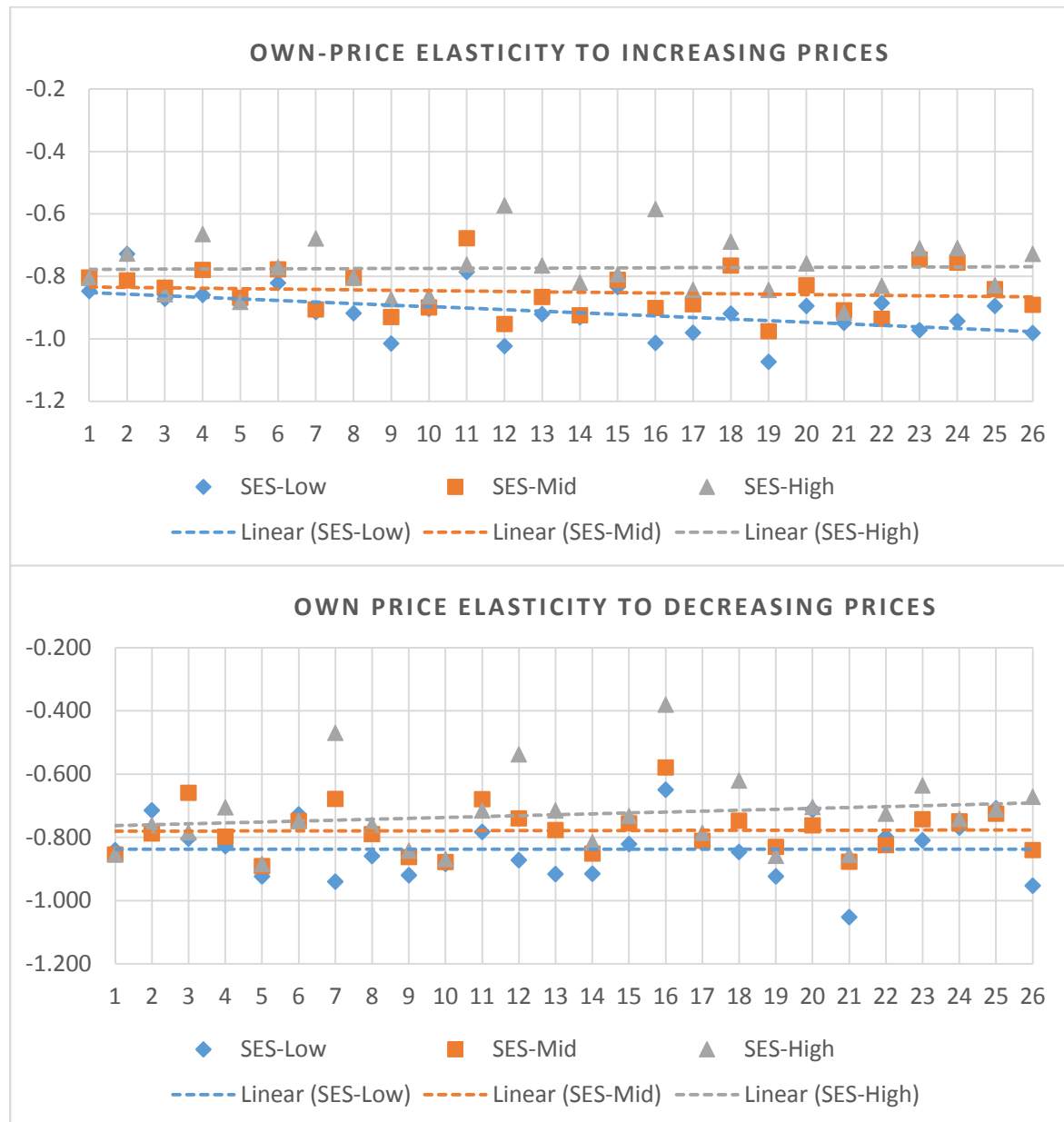
Notes:¹weighted to number of purchases; ²2.3g of sodium equivalent to <6g of salt; ³non-starch polysaccharides (NSP) fibre; ⁴ table salt accounts for 78% (0.44g) of sodium in the sodium content and table sugar accounts for 92% (13.3g) of sugar content in ‘other foods’.

Table 5. Change in energy and nutrient purchases per capita/day by SES if the price of all healthier foods decreased by 20% and the price of all less healthy foods increased by 20%

		Change in energy and nutrient content of daily per capita take-home purchases						Change in weekly household expenditure (£) ³
		Energy (kcal)	Sugar (g)	Sat. fat (g)	Sod. ¹ (g)	Prot. (g)	Fibre ² (g)	
Full sample	Price decrease of healthy foods	168.31	7.28	1.80	0.15	8.83	2.22	5.92
	Price increase of less healthy foods	-235.94	-13.61	-4.13	-0.36	-5.39	-1.30	-7.29
	<i>Difference</i>	-67.63	-6.33	-2.34	-0.21	3.44	0.92	+1.37
Low-SES	Decrease	161.18	6.15	1.55	0.16	9.37	2.21	5.67
	Increase	-252.47	-14.87	-4.21	-0.43	-6.14	-1.46	-6.62
	<i>Difference</i>	-91.29	-8.72	-2.65	-0.27	3.23	0.75	+0.95
Mid-SES	Decrease	172.24	7.36	1.85	0.14	8.49	2.27	5.77
	Increase	-227.05	-12.58	-4.08	-0.34	-5.24	-1.15	-7.22
	<i>Difference</i>	-54.81	-5.23	-2.23	-0.19	3.25	1.12	+1.45
High-SES	Decrease	169.29	9.45	1.83	0.17	9.13	2.22	6.53
	Increase	-241.10	-14.39	-4.29	-0.32	-5.17	-1.46	-7.61
	<i>Difference</i>	-71.80	-4.94	-2.46	-0.15	3.96	0.76	+1.08

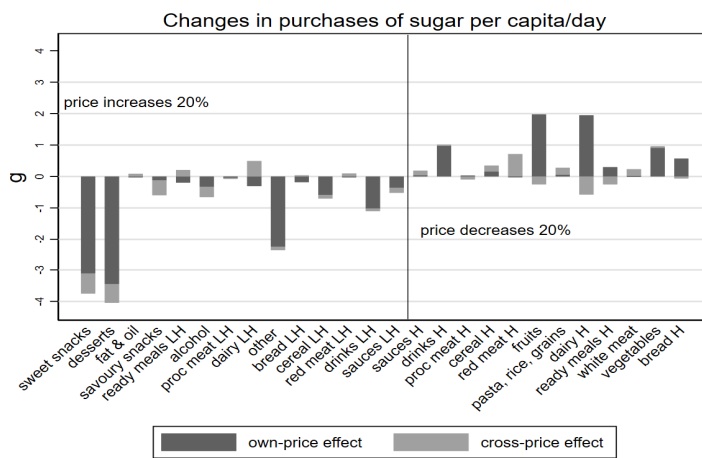
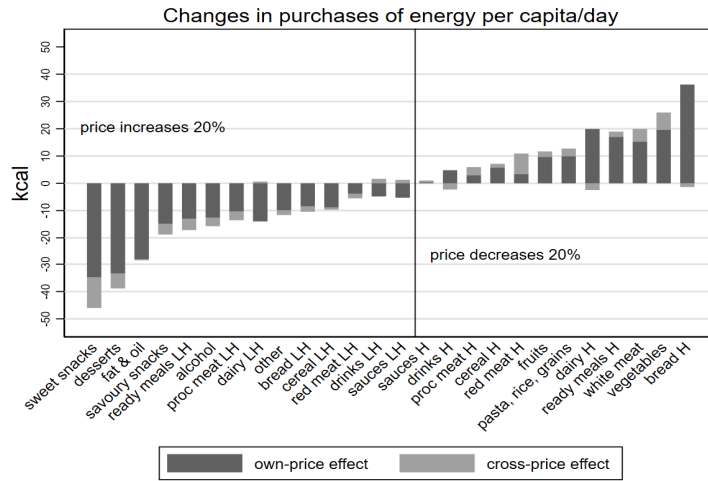
Notes: ¹2.3g of sodium equivalent to <6g of salt; ²non-starch polysaccharides (NSP) fibre; ³weighted expenditure at population level.

Figure 1. Own-price elasticity estimates by household SES, with food groups sorted by the average nutrient profile score (from left - healthier to right - less healthy).



Notes: Food groups on x-axis: 1 – vegetables (score -7.3); 2 – fruits (-4.2); 3 – cereal bars (healthier) (-2.9); 4 – bread&morning goods (healthier) (-1.6); 5 – non-alc drinks (healthier) (-1.3); 6 – pasta, grains, rice, dry pulses (-1.1); 7 – fresh&frozen read meat (healthier) (-0.4); 8 – ready meals&convenience (healthier) (-0.4); 9 – sauces&condiments (healthier)(-0.3); 10 – fresh&frozen white meat&fish, eggs (0.1); 11 – dairy (healthier)(0.1); 12 – processed meat&fish (healthier)(0.4); 13 – alcohol (0.4); 14 – non-alc drinks (less healthy) (3.3); 15 – desserts&puddings (9.5); 16 – fresh&frozen red meat (less healthy) (9.6); 17 – cereals&cereal bars (less healthy)(10.1); 18 – other foods (10.2); 19 – savoury snacks (10.7); 20 – bread&morning goods (less healthy) (10.8); 21 – ready meals&convenience (less healthy) (10.9); 22 – sauces&condiments (less healthy) (12.8); 23 – processed meat&fish (less healthy) (14.7); 24 – dairy (less healthy) (16.2); 25 – sweet snacks (19.7); 26 – fat&oil (21.0).

Figure 2. Changes in energy and nutrient content of take-home per capita daily purchases



*in 'other' group 92% of sugar purchases are from table sugar and other caloric sweeteners

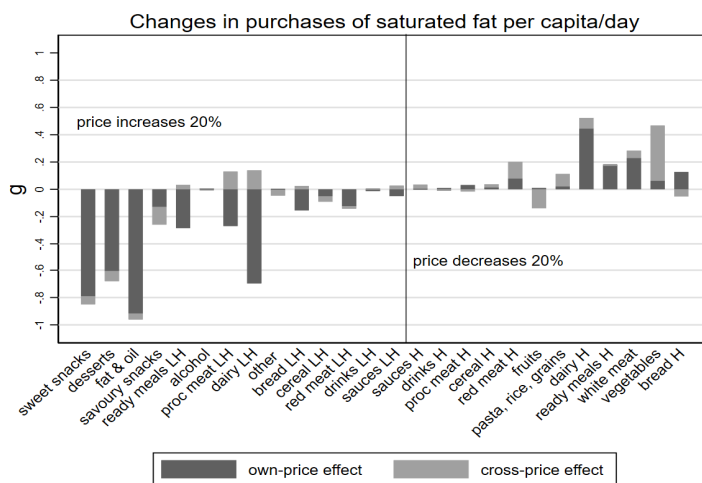
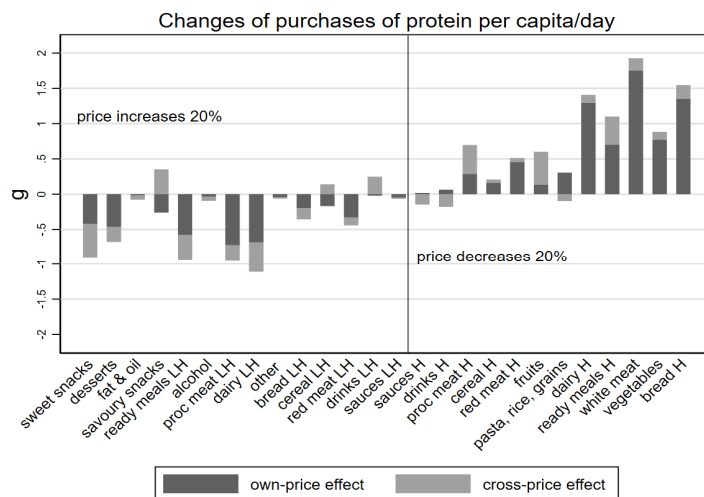
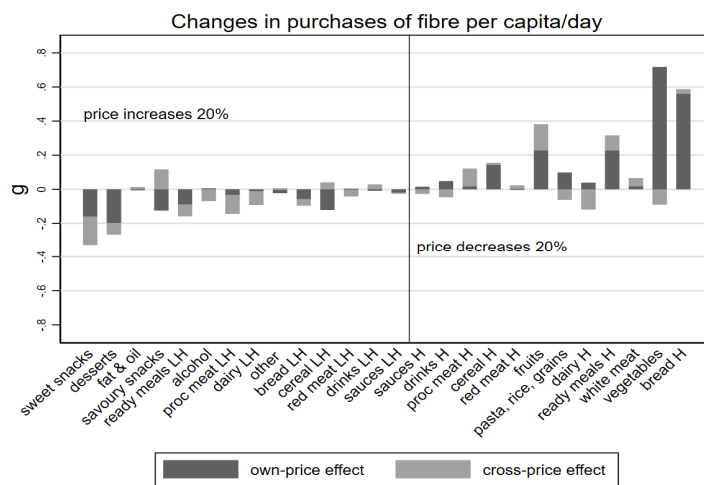
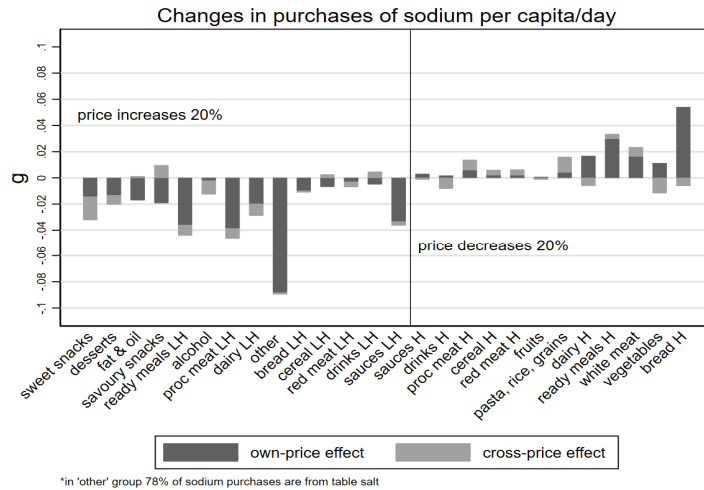


Fig.2 continued



Notes: Each bar on the figure represents the total change in energy or nutrient content of purchases due to price change in that particular food group only. See table 10 in supplementary material [INSERT LINK TO ONLINE FILE A] for standard errors and confidence intervals for total, own- and cross-price effects.

Fat tax or thin subsidy? How price increases and decreases affect the energy and nutrient content of food and beverage purchases in Great Britain

Highlights

- Energy and nutrient content of take-home purchases is above reference daily intake
- Demand for food is more responsive to price increases than to price decreases
- Price changes based on healthiness of food have a positive net effect on diet
- Price changes improve dietary quality of low-SES household food purchases most
- Greatest impact seen if price of sweet snacks, desserts, and fats/oils increases