

Difference in difference : Obesity

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November 5, 2021

1 Methodology

Data consists of an unbalanced panel of 128056 households for which we have data (at least) for the period spanning 2013 and the first 3 months of 2014. Our main variables are

1. Kcal of SD
2. Kcal of non-SD
3. Kcal of HCF
4. Kcal of non-HCF
5. Total taxable expenditure

Data is at week-individual level but we collapse it (by mean) at the monthly-individual level. We smooth all series using a MA with 3 lags, 2 forward terms and current observation; so that the smoother applied (by individual) is

$$(1/6)[x_{t-3} + x_{t-2} + x_{t-1} + x_t + x_{t+1} + x_{t+2}]$$

The first task is to define a pure treatment/control group. We will do this by choosing an *optimal* partition of the total taxable expenditure distribution, as high spenders will be more likely to be more sensitive to a price change in SD and HCF, therefore we will define this as the treatment group. Total taxable expenditure is defined as total expenditure in SD and HCF.

The *optimal* partition is found by solving the following problem

$$\begin{aligned}
& \min_{H,L} \sum_{t=-12}^{-2} |\beta_t| \\
& \text{s.t} \\
& (\beta_t)_{-12 \leq t \leq 12} = \operatorname{argmin} \left\{ \left(y_{it} - \sum_{k=-12}^{12} \alpha_k \mathbb{1}(t = k) - \sum_{k=-12}^{12} \beta_k \mathbb{1}(i = T, k = t) + \gamma \mathbb{1}(i = T) - \lambda_i \right)^2 \right\} \\
& T = \mathbb{1}(x_i \geq H) \\
& C = \mathbb{1}(x_i \leq L) \\
& \min(x_i) \leq L \leq H \leq \max(x_i)
\end{aligned}$$

Note that the coefficients β solve for a fixed effects regression including time calendar dummies and leads and lags in treatment effect¹ Moreover $(\beta_t)_{t=-12}^{-2}$ gives a ‘test’ on parallel trends, so what we are looking for is to find the optimal partition of Treatment/Control group so that a parallel trend is preserved. We do this in order to try to capture ‘true’ treatment effects.

Also note that in principle the partition is allowed to be non-symmetrical or to not span the whole distribution.

We use as a dependent variable (y) total calories of SD and HCF and total calories of SD. Variable (x_i) corresponds to total taxable expenditure of individual i . Finally, H and L are the respective cuts on the distribution to define Treatment/Control groups. Note that event study corresponds to $t = 0$.

¹As recommended by Borusyak and Jaravel (2016), but unlike McCrary (2007) and most event study papers, include all relative time dummies in the regression rather than “binning” periods below a or above b . Then we can just graph the periods from a to b if we want. But binning can cause bias if the trend isn’t flat for periods less than a or greater than b (Borusyak and Jaravel, 2016). Note that when there is no pure control group, binning periods less than a or greater than b (i.e. imposing flat trend for those periods) is needed to pin down calendar time fixed effects, which is why Borusyak and Jaravel (2016) recommend having a pure control, which pin down the calendar time fixed effects without having to make these additional assumptions.

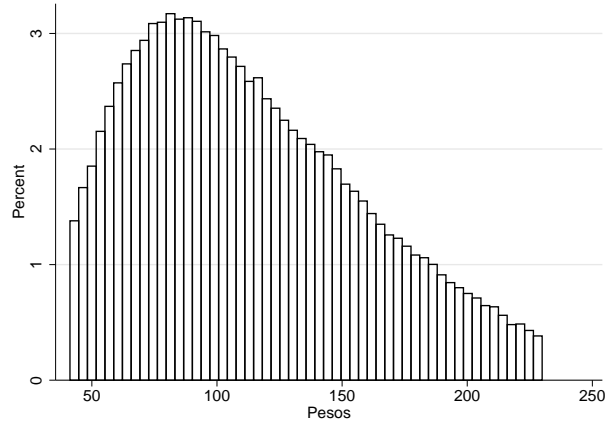
2 Results

Once we find the optimal “cuts” on the distribution of total taxable expenditure and define our pure Treatment/Control group, we graph

- (a) The average calorie consumption throughout time by treatment group
- (b) The coefficients of the fixed effect regression with leads and lags

The following graph shows the distribution (cut at the 95th percentile) of the total taxable expenditure.

Figure 1: Distribution taxable expenditure

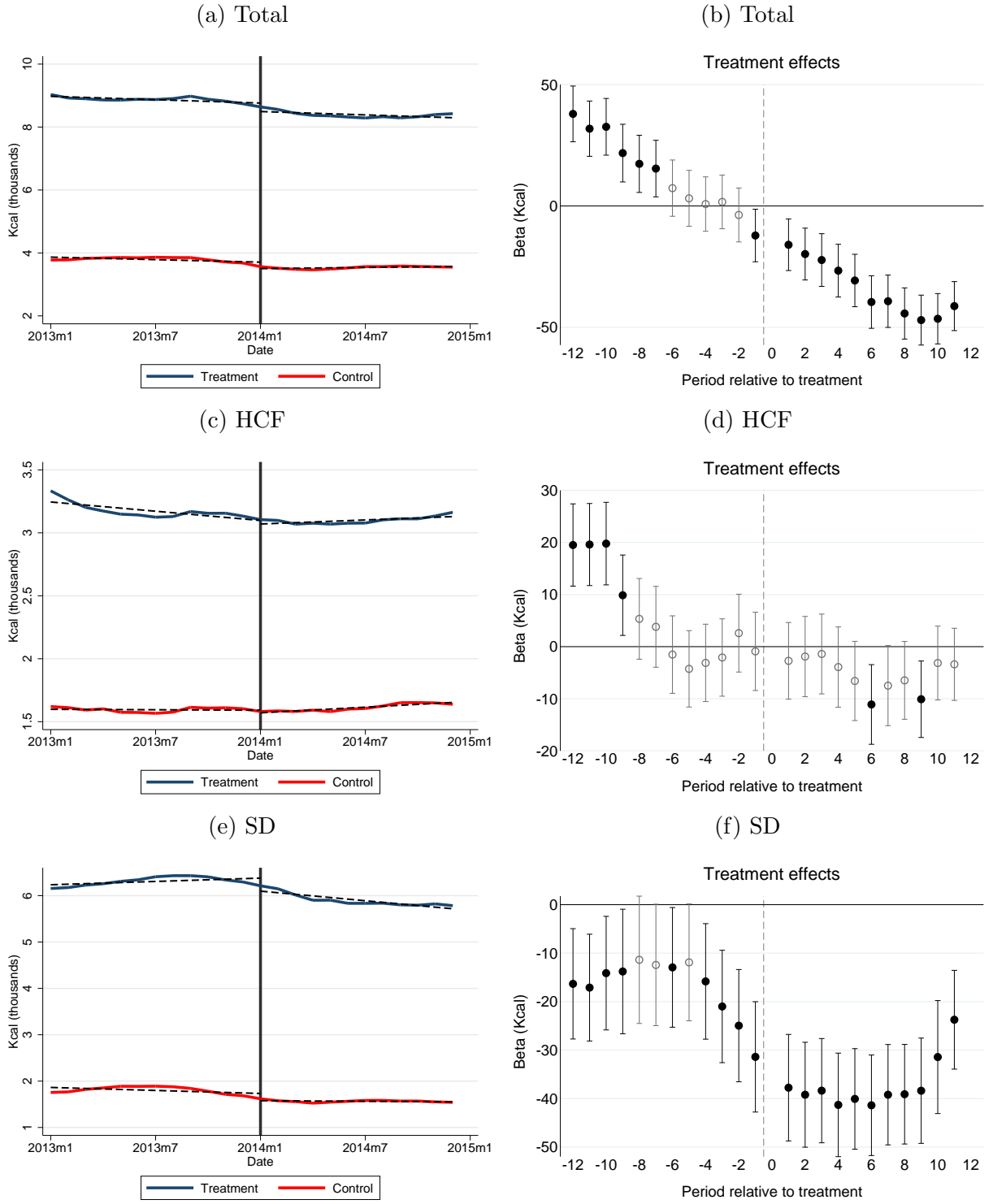


Notes: Do file: `dist_totalexpend.do`

The DiD specification is the following:

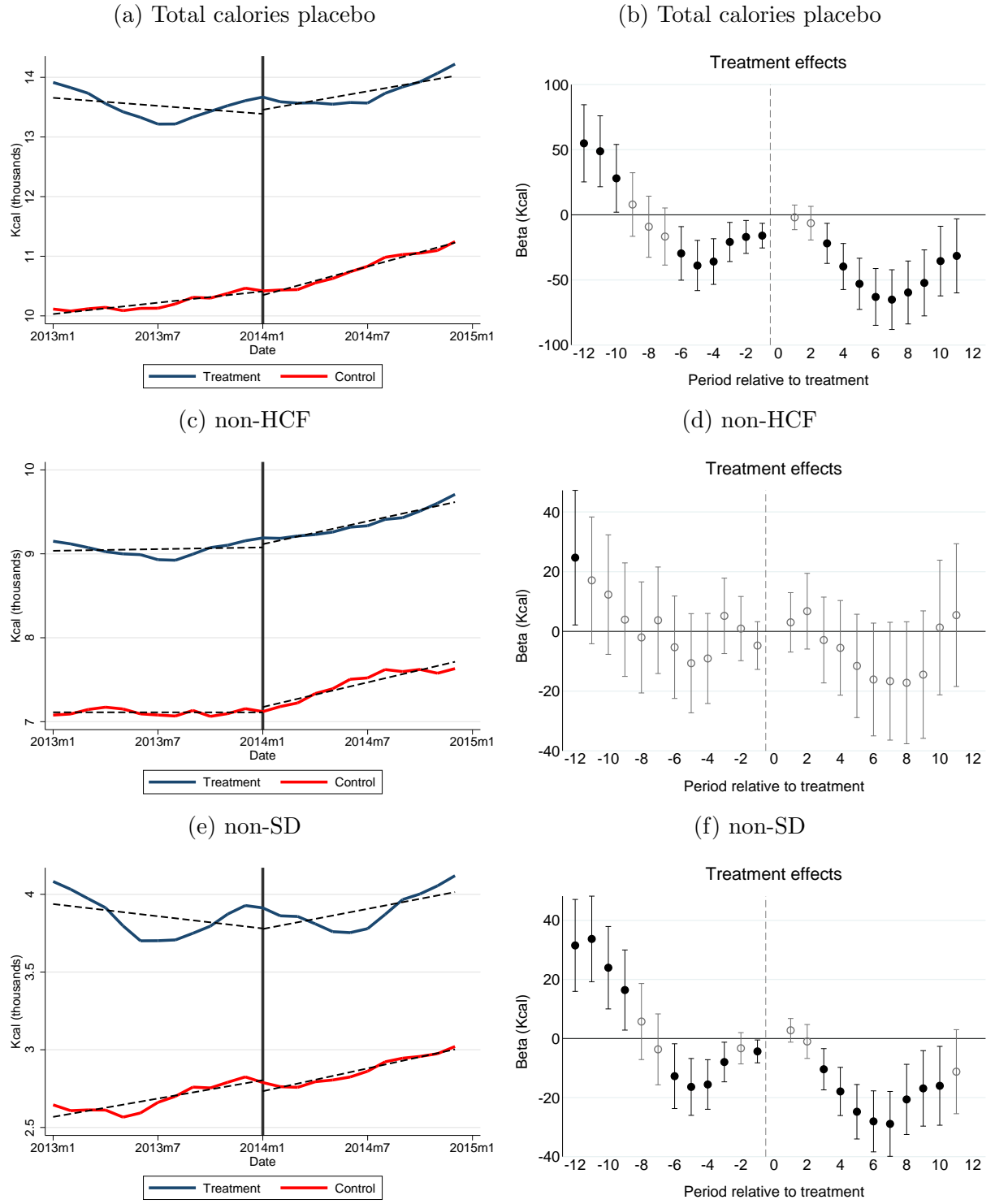
$$y_{it} = \sum_{k=-12}^{12} \alpha_k \mathbf{1}(t = k) + \sum_{k=-12}^{12} \beta_k \mathbf{1}(i = T, k = t) + \gamma \mathbf{1}(i = T) - \lambda_i + \epsilon_{it}$$

Figure 2: Treatment effects



Notes: Do file: did.do , beta_coef_did.do

Figure 3: Treatment effects - placebo



Notes: Do file: did.do , beta_coef_did.do