bp upf and na

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# BP and UPF and Na in NDNS Dissertation

# Method

## Introduction

This section takes the research question and explains how the data is used to answer the question.

There will be a description of the study and data collection. Then a section on governance and ethics in this project.

Data analysis starts with the relevant variables being identified and extracted. Some data may need to be recalculated or to be processed to make a more useable form. The population will be reviewed. Groups which bias the results are removed. Then there is a description of the data. The second analysis section compares the data between two cohorts. The third analysis section involves using linear regression to identify if there is a correlation between the BP and each of the key variables. Multivariable regression models are then generated. These models are finally examined to identify the relative importance of the different variables in developing an optimal model and what these models tell us about the relationship between our variables. A summary and conclusion will bring all these together.

## Research Question

What proportion of the association between blood pressure (SBP) and UPF intake can be explained by the changes in salt intake in England between 2008 and 2019?

The question can be split into parts, What was intake of salt between 2008 and 2019? What was intake of UPF between 2008 and 2019? What was BP between 2008 and 2019? Did each of these change over that time and how? Did the changes in any one affect any other? What are the sizes of the changes? Which element was most important in these changes?

All of these questions look for numbers as answers.

Answering the question starts with counting. The collected numbers are then compared in different ways to answer each part of the question.

## National Dietary and Nutritional Survey

This survey is a collaboration between government departments responsible for health and for food production. They have engaged academic partners to deliver reports on diet and nutrition across the United Kingdom. The study is designed to be representative across the whole area.

### Study design

This is a rolling cohort study which each year selects a new cohort of participants. The sample is approximately 1000 per year with 50% adults. The design has a random selection across postal units (psu). This is stratified to ensure a representative sample across the four nations and across regions within those countries. The sample is also representative for age and sex.

Having taken up the study, participants complete a 4 day food diary, and have an interview with a nurse which includes taking several measurements. Weighting is given for each annual survey to enable comparison across the years taking account for alterations in uptake and response completion.

### NDNS Dataset

The data from the NDNS study contains items about each individual,and their household. It contains a table with each item of food as recorded in their diary. There is a table with the overall intake of each of a large range of nutrients for the whole period. This is calculated from the diary using nutritional tables which are published as part of the dataset. The dataset is available via the UK national Data service for research purposes.

NDNS began before Monteiro’s processing based classification, Nova , was developed. There is no record of Nova food type in NDNS. This has been calculated from the food descriptions. I have used a table from Rauber et al. for Nova values in NDNS.

### University Research Governance and Ethical Review

The research has been carried out under the University governance. A proposal was discussed and agreed within the department. The need for ethical review was considered using the university research tool. The fact that the data is anonymised and there was no contact with participants means that there is minimal risk of harm to research participants. A certificate from the ethics department is in the appendix.

Other ethical issues include data custodianship ensuring that the the rights of the owners of the data and of the participants are still considered as part of the process of analysis and dissemination of the research.

Issues around the power structures which lead to privilege one research project or proposal over another are considered more in the positionality section.

## Data Processing

The storage of the data is in keeping with the research governance agreements of the University and the Data set owners. The data is read from its files using ‘r-studio’ with the processing being carried out using packages available from CRAN. I have used files which had been amalgamated into four batches. These are 2008-2012, 2013-2014, 2015-2016, 2017-2019.

Once the data labels are made consistent across the batches, weighting recalculation is done. This generates values which account for differences in population balance across the annual cohorts. These result from differences in compliance and uptake within and across the years.

The years are amalgamated and the nature of the variables is specified.

## Method for Data Analysis

### Descriptive data analysis

The data is summarised for the key continuous variables. The key variables are systolic BP (omsysval), UPF intake (Epcnt\_4) and Sodium intake (sodiummg). These variables are the ones which most relate to the research question.

There are a number of related variables in the dataset. These were chosen for relevance, reliability and practicality. These variables are ones which can also influence BP. They include Age, Sex, BMI, height and weight. Age at completion of education (educfinh), and IMD are also used. The population for early years are compared with those for later years.

The omsysval is a validated measurement with significant quality assessment within the dataset. Raw systolic BP values are present in the dataset but are made up of data with issues around quality. In particular the systolic BP values are assessed for the effects of exercise, temperature and ill health. The variable omsysval is a quality assured mean value which is reliable across the dataset.

The sodium value is one calculated from intake based on food diaries and standard food nutrient values. This only reflects standard foods and is the result of assumptions about the content being consistent. Serum sodium values are available for the early dataset, but not the later one. There are also values for 24 urinary sodium which is probably a better indicator of dietary sodium for parts of the dataset, but again these are not found in both time periods.

The food diaries need processing to identify the UPF intake. Each persons food diary entries are assessed against the Nova food classification from Rauber. Then the weight and energy content of the days food is calculated by Nova group. This is added to the intake for the other 3 days and the total intake by Nova group established.

The percentage of the total intake by weight (pcnt\_4) and by energy (Epcnt\_4)is then calculated for each of the 4 Nova categories. Nova group 4 or UPF intake is used for the study.

Mean values for the data are displayed with a comparison for weighted values. The exposure variables are sodium intake (Sodiummg), and ultra processed food intake (Epcnt\_4). The outcome variable, the mean systolic blood pressure (omsysval).

Key additional variables are considered in the later section looking at how they change across the populations. Age, sex, and bmi are important contributors. Education, NSSEC and IMD are also looked at.

### Exclusions

The relationship between salt and systolic blood pressure may be different in individuals with pathologically high BP. Those taking BP controlling medications may have a different relationship to sodium and UPF. These patients were excluded from the main analysis, however this affected the sample size and skewed the male female ratio. Analysis was done with exclusion and this produced results in line with those presented, but of smaller magnitude. This additional analysis is not presented here.

### Comparative Analysis

The second phase of analysis uses standard techniques to compare the means of the populations. A student’s t test is used as the populations are normally distributed continuous variables. The t tests compare the sodium, UPF, and systolic BP in the early cohort from the later. This will show how the intake and outcome have changed.

These are not the same participants so matched analysis, or time series analysis is not directly applicable.

Plots will be given to show the values in each of the available cohorts.

Other variables in the data are compared across to assess how the data changes. t tests are again used for continuous data, and chi squared tests are used for categorical data.

### Linear Regression

Analysis of the correlation between BP and sodium intake, and then BP and UPF intake is done using linear regression. This will give an indicator of the direction, and strength of any relationship between the variables. There is also anova analysis to understand the statistical significance of these results.

### Multiple Regression

Multivariable regression models are then developed to understand the interactions between variables and to develop a mathematical model of the relationship. The optimal model is one which best explains the pattern of data, but which also makes practical sense for the wider understanding of relationships. Assessment techniques try to understand the importance of including particular variables, and the form in which they are best included. Anova analysis here identifies how the addition of different variables changes the significance of other variables. This can suggest causative relationships.

### AIC and sensitivty Anaylsis

This section compares models side by side using assessment techniques to identify the best way of describing the data. The ‘best’ in part is determined by the whether a model is needed to predict more data, or just to understand the data available. Here it is about how best to describe the relationship between Na, UPF, and BP.

## Method Conclusion

This section has highlighted how the material for the study is brought together and how the governance and ethics fit with the data collection, processing and analysis to help us to derive the results which will be presented in the next section.

# Results

## Results Introduction

The results derive from the method outlined above and follow the pattern described. I will try to discuss the results having already described the method.

## Descriptive Data Analysis

This first table highlights the data from the years 2008-2012 at the beginning of the data collection.

## mean SE  
## Sodiummg 2181.9 16.835

## mean SE  
## omsysval 124.09 0.4614

## mean SE  
## Epcnt\_4 49.147 0.3052

## Sodiummg Epcnt\_4 omsysval   
## 2181.90505 49.14734 124.09433

## [1] "tbl\_svy" "survey.design2" "survey.design"

| **Characteristic** | **N = 652**1 |
| --- | --- |
| Sodium (mg) diet only | 2,139 (746) |
| Epcnt\_4 | 48 (14) |
| pcnt\_4 | 22 (14) |
| omsysval | 123 (13) |
| 1Mean (SD) | |

| **Characteristic** | **N = 831**1 |
| --- | --- |
| Sodiummg | 2,172 (715) |
| Epcnt\_4 | 47 (14) |
| pcnt\_4 | 22 (14) |
| omsysval | 124 (13) |
| 1Mean (SD) | |

The second table show the data from the second cohort 2017-19. The weighted values follow the unweighted/raw values.

## Sodiummg Epcnt\_4 omsysval   
## 1914.3144 45.9204 121.6101

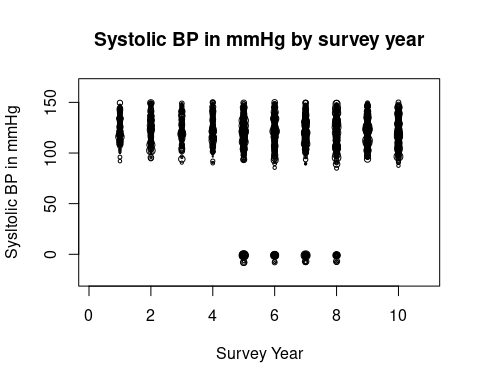
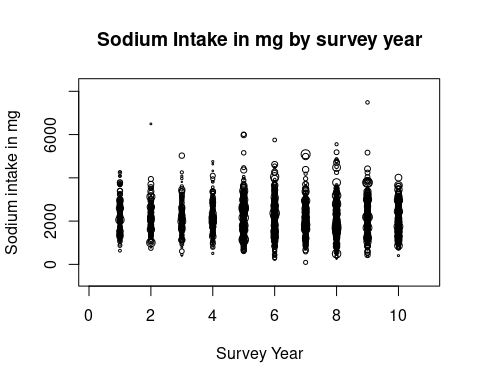
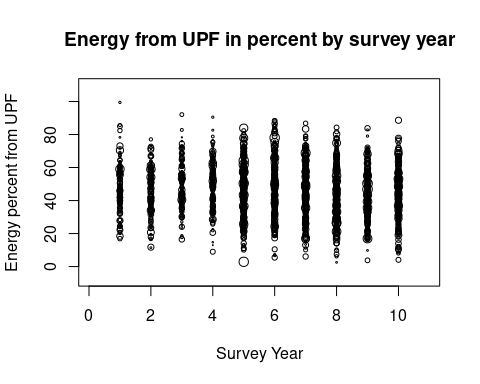
## Sodiummg Epcnt\_4 omsysval  
## means200812 2181.905 49.14734 124.0943  
## means201719 1914.314 45.92040 121.6101

## [1] "tbl\_svy" "survey.design2" "survey.design"

| **Characteristic** | **N = 592**1 |
| --- | --- |
| Sodium (mg) diet only | 1,995 (749) |
| Epcnt\_4 | 43 (14) |
| pcnt\_4 | 20 (15) |
| omsysval | 120 (13) |
| 1Mean (SD) | |

| **Characteristic** | **N = 1,150**1 |
| --- | --- |
| Sodiummg | 2,046 (782) |
| Epcnt\_4 | 43 (14) |
| pcnt\_4 | 21 (16) |
| omsysval | 120 (13) |
| 1Mean (SD) | |

These tables show that there has been a change between the two cohorts. They also show that the weighted samples are different to raw samples identifying the importance of analysing the data across weighted groups.

 These plots show how the percentage of energy derived from UPF, the sodium intake, and the Systolic bp have changed over the years. The graphs show that there is not a visible difference between the years. Statistical analysis will follow.

## Comparative analysis

### Comparison of key variables

comparing UPF and Sodium intake calculated from diet

In order to confirm there has been a change in intake a t.test compares the means of the two samples. One compares the means of sodium in years 1-4 with sodium in years 9-11.

The second compares the means of pcnt UPF intake in over the same periods. A third compares the percentage energy provided by UPF.

##   
## Design-based t-test  
##   
## data: Sodiummg ~ (SurveyYear >= 5)  
## t = -11.197, df = 971, p-value < 2.2e-16  
## alternative hypothesis: true difference in mean is not equal to 0  
## 95 percent confidence interval:  
## -268.3440 -188.3116  
## sample estimates:  
## difference in mean   
## -228.3278

##   
## Design-based t-test  
##   
## data: Epcnt\_4 ~ (SurveyYear >= 5)  
## t = -6.1092, df = 971, p-value = 1.447e-09  
## alternative hypothesis: true difference in mean is not equal to 0  
## 95 percent confidence interval:  
## -3.150439 -1.618544  
## sample estimates:  
## difference in mean   
## -2.384491

##   
## Design-based t-test  
##   
## data: EnergykJ\_4 ~ (SurveyYear >= 5)  
## t = -5.8284, df = 971, p-value = 7.61e-09  
## alternative hypothesis: true difference in mean is not equal to 0  
## 95 percent confidence interval:  
## -1363.5434 -676.6225  
## sample estimates:  
## difference in mean   
## -1020.083

## Var statistic p.value  
## 1: Epcnt\_4 -6.109 1.447e-09  
## 2: EkJ -5.828 7.610e-09  
## 3: Na -11.200 1.861e-27

It seems the mean percentage UPF intake changes from 48.8% to 59.2% energy and this increase is statistically significant. The mean sodium intake has changed from 2156.30 mg to 2574.33 mg and is also statistically significant with a p value less than 0.05.

what about outcome BP?

The next t tests compare mean systolic values in the two time periods and then the mean diastolic values.

## Var statistic p.value  
## 1: Sys -31.85 1.769e-147

##   
## Design-based t-test  
##   
## data: omsysval ~ SurveyYear >= 5  
## t = -31.855, df = 856, p-value < 2.2e-16  
## alternative hypothesis: true difference in mean is not equal to 0  
## 95 percent confidence interval:  
## -41.61183 -36.78157  
## sample estimates:  
## difference in mean   
## -39.1967

There is a change in mean systolic from 122-152 mmHg with a p value of 3.112e -7.

In summary there is statistically significant change in UPF and Na intake and also in both systolic and diastolic pressures.

Has another factor affected the BP change ?

### Comparative analysis of other variables

How are variables distributed between the two cohorts. The NDNS dataset was weighted to keep many of these the same between datasets. Continuous variables are assessed using ttests and categorical variables using chi squared tests to give p.values.

Age and Sex The age of the two datasets has changed but not in a statistically significant way.

## t   
## 0.3393779

## 2.5 % 97.5 %  
## SurveyYear >= 6TRUE -0.7994522 1.133785  
## attr(,"conf.level")  
## [1] 0.95

There is a statistically significant change in the sex distribution of the two groups.

## SurveyYear >= 5  
## Sex FALSE TRUE  
## Male 2796.205 4903.148  
## Female 2896.522 5059.125

## SurveyYear  
## Sex 1 2 3 4 5 6 7  
## Male 50.06198 47.96937 32.01676 44.04683 228.00517 172.35312 180.69957  
## Female 70.41059 84.51633 74.71734 90.27609 230.15312 187.31313 158.33500  
## SurveyYear  
## Sex 8 9 10  
## Male 177.33201 157.79591 167.49868  
## Female 210.83607 155.19802 203.46491

## X-squared   
## 0.002174653

This might be due to differences in the numbers of excluded participants. In particular there may be more younger people and women taking e.g. bblockers in one group.

This table suggests that there is a significant difference between the bmi of the groups.

## t   
## -5.832169

## 2.5 % 97.5 %  
## SurveyYear >= 5TRUE -1.218884 -0.605134  
## attr(,"conf.level")  
## [1] 0.95

There is a difference in the age of finishing education.

## SurveyYear >= 5  
## educfinh FALSE TRUE  
## 1 293.32724 81.99285  
## 2 11.86102 29.20814  
## 3 270.77232 232.97656  
## 4 773.87207 999.48053  
## 5 1169.16847 2313.79944  
## 6 399.62423 674.81579  
## 7 548.19728 1039.69916  
## 8 1247.26251 2924.32032

## F   
## 3.507224

## X-squared   
## 1.859558e-16

## SurveyYear >= 5  
## EIMD\_2010\_quintile FALSE TRUE  
## 1 1046.2033 1760.1701  
## 2 1013.0021 1714.6541  
## 3 875.8083 1487.7434  
## 4 887.7392 1754.3622  
## 5 938.0365 1555.3653

## SurveyYear  
## EIMD\_2010\_quintile 1 2 3 4 5 6  
## 1 311.0077 358.0844 303.2495 284.3305 356.9996 270.6352  
## 2 376.0997 266.2440 249.9055 324.5423 340.0152 290.6916  
## 3 229.2665 221.1998 261.7770 339.7547 252.0957 211.1799  
## 4 220.0337 290.2782 276.0861 279.9311 301.9044 272.4083  
## 5 340.2418 275.8057 283.7281 226.9693 287.9251 286.4669  
## SurveyYear  
## EIMD\_2010\_quintile 7 8 9 10 11  
## 1 317.9274 243.6063 270.3710 356.0118 298.7189  
## 2 245.0458 287.7351 341.5636 299.9665 254.5798  
## 3 269.8687 274.9745 232.4486 226.1737 320.2973  
## 4 370.0786 301.9084 311.4369 307.2442 242.3131  
## 5 258.9584 277.5179 205.9343 309.9395 241.5220

## X-squared   
## 0.2207905

The differences in qimd, are not statistically significant.

These values identify a significant difference in the number of vegetarians

## SurveyYear >= 5  
## vegetarn FALSE TRUE  
## vegetarian 127.119927 287.261777  
## vegan 4.833554 39.303208  
## not vegetarian 5560.773794 9635.336015

## SurveyYear  
## vegetarn 1 2 3 4  
## vegetarian 28.1016898 32.6384769 18.5134165 47.8693619  
## vegan 0.3054694 3.6842181 0.6466089 0.1973727  
## not vegetarian 1430.8814086 1393.0175719 1353.1444815 1383.8623749  
## SurveyYear  
## vegetarn 5 6 7 8  
## vegetarian 34.9993979 33.7237598 37.6135652 43.2703428  
## vegan 2.3917728 0.0000000 2.0005228 1.6529873  
## not vegetarian 1446.8321137 1328.1124412 1401.9338044 1359.9600030  
## SurveyYear  
## vegetarn 9 10 11  
## vegetarian 42.4907099 50.8124409 44.3583815  
## vegan 11.0545087 6.6177611 15.5865888  
## not vegetarian 1390.5301357 1423.3786493 1284.8176624

## X-squared   
## 0.02450092

## Regression Analysis

#### Linear Regression

Simple linear regression equations look for the relationship between the dependant variable, and the independent variable. For these I am looking at the whole dataset

The regression models are examined for Sodium and UP against BP. These use the populations where participants have been excluded. (analysis including these makes no difference!!)

omsysval is compared to EnergykJ

## (Intercept) Epcnt\_4   
## 109.4493819 -0.2724466

then sodiummg

## (Intercept) Sodiummg   
## 85.078412191 0.005582757

##   
## % Table created by stargazer v.5.2.3 by Marek Hlavac, Social Policy Institute. E-mail: marek.hlavac at gmail.com  
## % Date and time: Thu, Mar 30, 2023 - 17:10:58  
## % Requires LaTeX packages: dcolumn   
## \begin{table}[!htbp] \centering   
## \caption{Results}   
## \label{}   
## \begin{tabular}{@{\extracolsep{5pt}}lD{.}{.}{-3} D{.}{.}{-3} }   
## \\[-1.8ex]\hline   
## \hline \\[-1.8ex]   
## & \multicolumn{2}{c}{\textit{Dependent variable:}} \\   
## \cline{2-3}   
## \\[-1.8ex] & \multicolumn{2}{c}{omsysval} \\   
## \\[-1.8ex] & \multicolumn{1}{c}{(1)} & \multicolumn{1}{c}{(2)}\\   
## \hline \\[-1.8ex]   
## Sodiummg & 0.006^{\*\*\*} & \\   
## & (0.001) & \\   
## & & \\   
## Epcnt\\_4 & & -0.272^{\*\*\*} \\   
## & & (0.049) \\   
## & & \\   
## Constant & 85.078^{\*\*\*} & 109.449^{\*\*\*} \\   
## & (2.268) & (2.539) \\   
## & & \\   
## \hline \\[-1.8ex]   
## Observations & \multicolumn{1}{c}{10,107} & \multicolumn{1}{c}{10,107} \\   
## Log Likelihood & \multicolumn{1}{c}{-57,132.200} & \multicolumn{1}{c}{-57,137.170} \\   
## Akaike Inf. Crit. & \multicolumn{1}{c}{114,268.400} & \multicolumn{1}{c}{114,278.300} \\   
## \hline   
## \hline \\[-1.8ex]   
## \textit{Note:} & \multicolumn{2}{r}{$^{\*}$p$<$0.1; $^{\*\*}$p$<$0.05; $^{\*\*\*}$p$<$0.01} \\   
## \end{tabular}   
## \end{table}

There are relationships between Na and g pcnt as well as E pcnt and omsysval .

In conclusion the linear regression models show that there are correlations between the systolic BP and each of the key variables.

#### multi variable regression

This uses a model of variables. It can highlight the contributions of each variable. The intention is to develop an optimal model which mathematically describes the situation.

In particular the research question asks about the relationship between Sodium and UPF intake with BP. The models will reflect this question with models looking to include or exclude particular variables. Comparisons between these models are then made using sensitivity analysis, identifying how sensitive the model is to sodium, or other factors

This first model looks at the relationships between BP and Age and Sex education and IMD all of which may have an effect on BP. This model excludes UPF and Na.

## (Intercept) Age SexFemale bmival   
## 63.8629864 0.5684534 -6.1557760 1.4457394   
## educfinh2 educfinh3 educfinh4 educfinh5   
## 1.9233972 -23.0251651 -22.6197827 -21.2258264   
## educfinh6 educfinh7 educfinh8 EIMD\_2010\_quintile2   
## -29.1864363 -18.3646005 -18.7861647 1.8093644   
## EIMD\_2010\_quintile3 EIMD\_2010\_quintile4 EIMD\_2010\_quintile5   
## -0.9827437 -4.9714160 -3.9077475

## Anova table: (Rao-Scott LRT)  
## svyglm(formula = omsysval ~ Age, design = ndns\_1\_11d, na.action = na.exclude)  
## stats DEff df ddf p   
## Age 1769230 5152.1 1.0 856 < 2.2e-16 \*\*\*  
## Sex 66818 5415.5 1.0 855 0.0004998 \*\*\*  
## bmival 1967932 7601.7 1.0 848 < 2.2e-16 \*\*\*  
## educfinh 6865959 4810.8 7.0 741 < 2.2e-16 \*\*\*  
## EIMD\_2010\_quintile 3619401 6794.8 4.0 429 < 2.2e-16 \*\*\*  
## ---  
## Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

## [1] 48008.04

This first model shows that all these variables, Age, Sex, education, IMD, and bmi, give statistically significant coefficients for the model which suggests that they do have an important part to play in any optimal model.

The next model adds Sodiummg.

## (Intercept) Age SexFemale Sodiummg   
## 57.676434338 0.570173655 -4.612795500 0.002946435   
## bmival educfinh2 educfinh3 educfinh4   
## 1.406514873 3.624974944 -21.989209987 -22.047253901   
## educfinh5 educfinh6 educfinh7 educfinh8   
## -20.906646163 -28.635031126 -18.152873105 -18.510535323   
## EIMD\_2010\_quintile2 EIMD\_2010\_quintile3 EIMD\_2010\_quintile4 EIMD\_2010\_quintile5   
## 1.693260342 -1.376257736 -5.140956060 -3.964198615

## Anova table: (Rao-Scott LRT)  
## svyglm(formula = omsysval ~ Age, design = ndns\_1\_11d, na.action = na.exclude)  
## stats DEff df ddf p   
## Age 1769230 5152.1 1.0 856 < 2.2e-16 \*\*\*  
## Sex 66818 5415.5 1.0 855 0.0004998 \*\*\*  
## Sodiummg 168043 7128.3 1.0 854 1.56e-06 \*\*\*  
## bmival 1860042 7607.8 1.0 847 < 2.2e-16 \*\*\*  
## educfinh 6862697 4779.8 7.0 740 < 2.2e-16 \*\*\*  
## EIMD\_2010\_quintile 3597547 6707.6 4.0 428 < 2.2e-16 \*\*\*  
## ---  
## Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

This second model gives Sodiummg, educfinh, and IMD statistical significance. VitaminD shows no statistical significance, TotalEMJ and sqrt(pcnt) and ethgrp2 all have limited significance.

Now we add UPF as total energy from nova 4 or UPF

## (Intercept) Age SexFemale Sodiummg   
## 5.624409e+01 5.803896e-01 -4.412730e+00 2.104280e-03   
## EnergykJ\_4 bmival educfinh2 educfinh3   
## 1.662087e-04 1.409758e+00 3.957273e+00 -2.195893e+01   
## educfinh4 educfinh5 educfinh6 educfinh7   
## -2.197925e+01 -2.076418e+01 -2.852096e+01 -1.784456e+01   
## educfinh8 EIMD\_2010\_quintile2 EIMD\_2010\_quintile3 EIMD\_2010\_quintile4   
## -1.817195e+01 1.713897e+00 -1.329809e+00 -5.129742e+00   
## EIMD\_2010\_quintile5   
## -3.962486e+00

## Anova table: (Rao-Scott LRT)  
## svyglm(formula = omsysval ~ Age, design = ndns\_1\_11d, na.action = na.exclude)  
## stats DEff df ddf p   
## Age 1769230 5152.1 1.0 856 < 2.2e-16 \*\*\*  
## Sex 66818 5415.5 1.0 855 0.0004998 \*\*\*  
## Sodiummg 168043 7128.3 1.0 854 1.56e-06 \*\*\*  
## EnergykJ\_4 10540 6501.8 1.0 853 0.2058373   
## bmival 1866294 7578.6 1.0 846 < 2.2e-16 \*\*\*  
## educfinh 6854647 4774.7 7.0 739 < 2.2e-16 \*\*\*  
## EIMD\_2010\_quintile 3593502 6701.1 4.0 427 < 2.2e-16 \*\*\*  
## ---  
## Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

UPF does not seem significant…

but when removing sodiummg

## (Intercept) Age SexFemale EnergykJ\_4   
## 5.763427e+01 5.892916e-01 -4.908157e+00 3.234609e-04   
## bmival educfinh2 educfinh3 educfinh4   
## 1.430233e+00 3.516575e+00 -2.238999e+01 -2.216897e+01   
## educfinh5 educfinh6 educfinh7 educfinh8   
## -2.077102e+01 -2.865772e+01 -1.764682e+01 -1.797393e+01   
## EIMD\_2010\_quintile2 EIMD\_2010\_quintile3 EIMD\_2010\_quintile4 EIMD\_2010\_quintile5   
## 1.784944e+00 -1.111238e+00 -5.043898e+00 -3.935814e+00

## Anova table: (Rao-Scott LRT)  
## svyglm(formula = omsysval ~ Age, design = ndns\_1\_11d, na.action = na.exclude)  
## stats DEff df ddf p   
## Age 1769230 5152.1 1.0 856 < 2.2e-16 \*\*\*  
## Sex 66818 5415.5 1.0 855 0.0004998 \*\*\*  
## EnergykJ\_4 108159 6443.9 1.0 854 4.954e-05 \*\*\*  
## bmival 1922691 7598.4 1.0 847 < 2.2e-16 \*\*\*  
## educfinh 6850386 4791.3 7.0 740 < 2.2e-16 \*\*\*  
## EIMD\_2010\_quintile 3600686 6759.3 4.0 428 < 2.2e-16 \*\*\*  
## ---  
## Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

the UPF becomes significant! This suggests that the effect of UPF is mediated by Sodium!!

comparing AIC for these three models

lmM$aic

## [1] 48000.08

lmMna$aic

## [1] 48001.91

lmM2$aic

## [1] 48000.75

#AIC(lmM,lmMna,lmM2)  
sense1 <- (lm1AS$aic - lmM2$aic)\*100/lm1AS$aic   
sense1

## [1] 0.01517998

we find that the lowest AIC is given by the model without UPF!! Though all the models with UPF have a lower aic than the model without.

## Summary of Results

There is a table with summary values for the key variables across the dataset.

Statistical analysis of the key variables shows the change in all the variables between the two time periods.

Confounding variables are analysed and show if there has been a significant change in the balance of the populations.

Regression shows a degree of association between the BP and UPF intake by weight and by energy. It also shows the same for sodium intake.

Using Anova analysis of different multi variable regression models the key variables are significant for sodium in several models, and sometimes for UPF.

## Conclusion

The percentage by weight of NOVA group 4 foods increased from 2008 to 2019. The percentage by energy of NOVA group 4 foods increased from 2008 to 2019. The mean sodium intake in mg increased between the two time periods. The systolic and diastolic BP have increased between the two time periods.

In each period there is a correlation between systolic BP and sodium intake. In each period there is a correlation between systolic BP and UPF intake.

The regression models identify that age and sex are statistically significant contributors to the BP.

The regression models identify that sodium intake is an important contributor to any optimal model. That when sodium intake is present UPF intake is no longer significant, but still has some effect.