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. A summary and conclusion will bring all these together.

## Research Question

What proportion of the association between blood pressure (SBP/DBP) 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., but also one from Colombet (personal communication)

### 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.

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, with mean, median, and range 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 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 years 1-4 are compared with those for years 9-11.

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 are 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. Analysis was done with them included and this produced results in line with those presented, but of greater 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 teh 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 eleven 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 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.

### 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

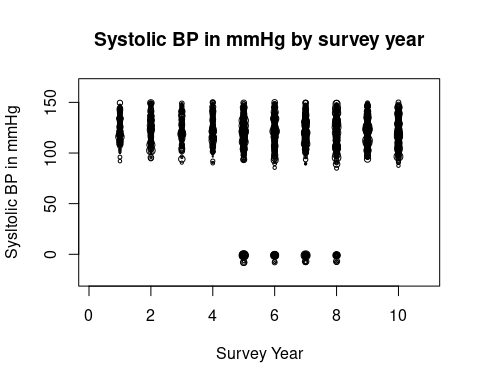
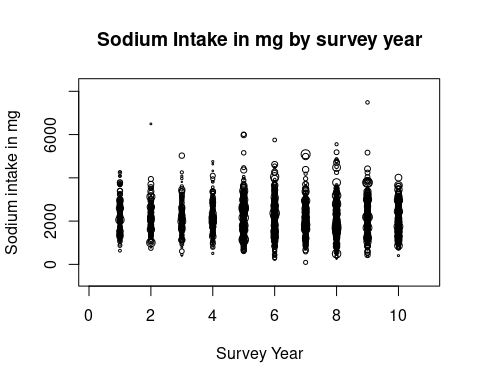
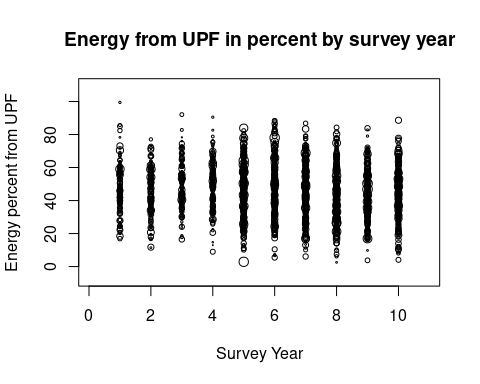
## Sodiummg Epcnt\_4 omsysval  
## means200812 2181.905 49.14734 124.0943  
## means201719 1914.314 45.92040 121.6101

## [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 = -1.798, df = 533, p-value = 0.07274  
## alternative hypothesis: true difference in mean is not equal to 0  
## 95 percent confidence interval:  
## -160.020319 7.077685  
## sample estimates:  
## difference in mean   
## -76.47132

##   
## Design-based t-test  
##   
## data: Epcnt\_4 ~ (SurveyYear >= 5)  
## t = -3.1937, df = 533, p-value = 0.001488  
## alternative hypothesis: true difference in mean is not equal to 0  
## 95 percent confidence interval:  
## -4.290859 -1.022575  
## sample estimates:  
## difference in mean   
## -2.656717

##   
## Design-based t-test  
##   
## data: EnergykJ\_4 ~ (SurveyYear >= 5)  
## t = 0.25042, df = 533, p-value = 0.8024  
## alternative hypothesis: true difference in mean is not equal to 0  
## 95 percent confidence interval:  
## -592.1272 765.1474  
## sample estimates:  
## difference in mean   
## 86.51013

## Var statistic p.value  
## 1: Epcnt\_4 -3.1940 0.001488  
## 2: EkJ 0.2504 0.802400  
## 3: Na -1.7980 0.072740

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 -13.11 2.971e-34

##   
## Design-based t-test  
##   
## data: omsysval ~ SurveyYear >= 5  
## t = -13.115, df = 533, p-value < 2.2e-16  
## alternative hypothesis: true difference in mean is not equal to 0  
## 95 percent confidence interval:  
## -21.76629 -16.09515  
## sample estimates:  
## difference in mean   
## -18.93072

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   
## -3.029559

## 2.5 % 97.5 %  
## SurveyYear >= 6TRUE -4.473062 -0.9540307  
## 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 292.7118 1822.0360  
## Female 537.8931 1925.6327

## 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   
## -3.878727

## 2.5 % 97.5 %  
## SurveyYear >= 5TRUE -1.916213 -0.6277231  
## attr(,"conf.level")  
## [1] 0.95

There is a difference in the age of finishing education.

## SurveyYear >= 5  
## educfinh FALSE TRUE  
## 1 23.45204 50.28225  
## 2 0.00000 10.89498  
## 3 27.28979 46.84485  
## 4 158.49005 300.15352  
## 5 211.63321 861.24032  
## 6 81.30505 199.83061  
## 7 94.79068 411.44564  
## 8 233.64408 1051.93547

## F   
## 1.659393

## X-squared   
## 0.003125183

## SurveyYear >= 5  
## EIMD\_2010\_quintile FALSE TRUE  
## 1 203.9000 667.3112  
## 2 144.9402 700.1465  
## 3 120.7524 575.2667  
## 4 119.1947 635.9267  
## 5 116.0614 571.3128

## SurveyYear  
## EIMD\_2010\_quintile 1 2 3 4 5  
## 1 34.450634 43.081972 30.332980 36.167257 107.210530  
## 2 42.001338 15.543889 19.379452 25.459584 112.881362  
## 3 6.718736 24.981581 18.047432 35.550539 76.426839  
## 4 20.972432 24.586130 17.529068 21.110252 86.366120  
## 5 18.279286 26.998083 17.859308 18.847907 95.658774  
## SurveyYear  
## EIMD\_2010\_quintile 6 7 8 9 10  
## 1 67.563764 69.476533 65.589014 70.384572 91.157343  
## 2 88.956205 61.745950 80.788989 71.459717 78.744100  
## 3 61.295359 71.839752 81.801002 53.849555 61.149964  
## 4 67.666285 85.070207 76.370503 59.310644 74.428350  
## 5 66.718229 52.236115 72.709665 49.016260 67.230437

## X-squared   
## 0.5773836

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 10.91746 126.62985  
## vegan 0.00000 19.83417  
## not vegetarian 819.68744 3601.20467

## SurveyYear  
## vegetarn 1 2 3 4 5  
## vegetarian 0.4237189 0.3356719 1.2994000 4.4345398 16.4932611  
## vegan 0.0000000 0.0000000 0.0000000 0.0000000 1.3285491  
## not vegetarian 120.0488463 132.1500282 105.4347022 129.8883802 440.3364792  
## SurveyYear  
## vegetarn 6 7 8 9 10  
## vegetarian 11.6634214 5.4696643 12.9289343 9.8631868 18.8966220  
## vegan 0.0000000 1.1898140 0.9294605 6.5136847 1.8351724  
## not vegetarian 348.0028211 332.3750889 374.3096934 296.6170594 350.2317998

## X-squared   
## 0.05068419

## 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) EnergykJ\_4   
## 1.111762e+02 -2.157041e-04

then sodiummg

## (Intercept) Sodiummg   
## 111.957112528 -0.001816932

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   
## 85.8191145 0.3083563 -2.7207680 0.6405843   
## educfinh2 educfinh3 educfinh4 educfinh5   
## 18.9841272 -2.2688388 -1.8389628 -3.7693031   
## educfinh6 educfinh7 educfinh8 EIMD\_2010\_quintile2   
## -9.6013030 -2.4369487 -1.3725591 -2.0401996   
## EIMD\_2010\_quintile3 EIMD\_2010\_quintile4 EIMD\_2010\_quintile5   
## -1.9970928 -4.1452849 -5.4437472

## Anova table: (Rao-Scott LRT)  
## svyglm(formula = omsysval ~ Age, design = ndns\_1\_11ed, na.action = na.exclude)  
## stats DEff df ddf p   
## Age 57478.6 2392.5 1.0 533 1.380e-06 \*\*\*  
## Sex 5084.8 2683.0 1.0 532 0.1719   
## bmival 73384.6 3222.5 1.0 516 2.592e-06 \*\*\*  
## educfinh 1192051.5 1924.3 7.0 440 < 2.2e-16 \*\*\*  
## EIMD\_2010\_quintile 503047.4 2556.3 4.0 279 < 2.2e-16 \*\*\*  
## ---  
## Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

## [1] 14303.28

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   
## 89.076363767 0.298018706 -3.426440252 -0.001237575   
## bmival educfinh2 educfinh3 educfinh4   
## 0.642400711 18.625586950 -2.533634585 -1.700732289   
## educfinh5 educfinh6 educfinh7 educfinh8   
## -3.581066217 -9.584009469 -2.213865629 -1.262891941   
## EIMD\_2010\_quintile2 EIMD\_2010\_quintile3 EIMD\_2010\_quintile4 EIMD\_2010\_quintile5   
## -2.007001284 -1.843283610 -4.102288897 -5.527442694

## Anova table: (Rao-Scott LRT)  
## svyglm(formula = omsysval ~ Age, design = ndns\_1\_11ed, na.action = na.exclude)  
## stats DEff df ddf p   
## Age 57478.6 2392.5 1.0 533 1.380e-06 \*\*\*  
## Sex 5084.8 2683.0 1.0 532 0.1719   
## Sodiummg 4138.9 3754.6 1.0 531 0.2959   
## bmival 74949.8 3205.3 1.0 515 1.915e-06 \*\*\*  
## educfinh 1187410.8 1936.1 7.0 439 < 2.2e-16 \*\*\*  
## EIMD\_2010\_quintile 503387.4 2573.1 4.0 278 < 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   
## 8.858537e+01 3.013972e-01 -3.357285e+00 -1.548927e-03   
## EnergykJ\_4 bmival educfinh2 educfinh3   
## 5.930366e-05 6.440246e-01 1.885612e+01 -2.564861e+00   
## educfinh4 educfinh5 educfinh6 educfinh7   
## -1.655187e+00 -3.530697e+00 -9.519688e+00 -2.095537e+00   
## educfinh8 EIMD\_2010\_quintile2 EIMD\_2010\_quintile3 EIMD\_2010\_quintile4   
## -1.122974e+00 -2.024350e+00 -1.811938e+00 -4.121010e+00   
## EIMD\_2010\_quintile5   
## -5.547742e+00

## Anova table: (Rao-Scott LRT)  
## svyglm(formula = omsysval ~ Age, design = ndns\_1\_11ed, na.action = na.exclude)  
## stats DEff df ddf p   
## Age 57478.64 2392.50 1.00 533 1.380e-06 \*\*\*  
## Sex 5084.77 2683.00 1.00 532 0.1719   
## Sodiummg 4138.89 3754.60 1.00 531 0.2959   
## EnergykJ\_4 276.15 2939.90 1.00 530 0.7528   
## bmival 74994.97 3206.10 1.00 514 1.908e-06 \*\*\*  
## educfinh 1187565.88 1950.40 7.00 438 < 2.2e-16 \*\*\*  
## EIMD\_2010\_quintile 503072.17 2565.00 4.00 277 < 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   
## 8.682514e+01 3.037660e-01 -2.910149e+00 -4.552691e-05   
## bmival educfinh2 educfinh3 educfinh4   
## 6.396885e-01 1.873790e+01 -2.296008e+00 -1.847230e+00   
## educfinh5 educfinh6 educfinh7 educfinh8   
## -3.771615e+00 -9.647342e+00 -2.484703e+00 -1.458792e+00   
## EIMD\_2010\_quintile2 EIMD\_2010\_quintile3 EIMD\_2010\_quintile4 EIMD\_2010\_quintile5   
## -2.020469e+00 -1.991450e+00 -4.122609e+00 -5.444328e+00

## Anova table: (Rao-Scott LRT)  
## svyglm(formula = omsysval ~ Age, design = ndns\_1\_11ed, na.action = na.exclude)  
## stats DEff df ddf p   
## Age 57478.6 2392.5 1.0 533 1.38e-06 \*\*\*  
## Sex 5084.8 2683.0 1.0 532 0.1719   
## EnergykJ\_4 2545.0 3770.4 1.0 531 0.4114   
## bmival 74185.1 3212.3 1.0 515 2.21e-06 \*\*\*  
## educfinh 1190031.4 1929.2 7.0 439 < 2.2e-16 \*\*\*  
## EIMD\_2010\_quintile 501869.5 2556.2 4.0 278 < 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] 14304.48

lmMna$aic

## [1] 14305.2

lmM2$aic

## [1] 14306.39

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

## [1] -0.02169342

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.