

FRS workshop

September 2022

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1 Benefit Comparison – Year: 2020-21

1.1 Create Both1

Create a new dataset called ‘both1’ from merging the FRS adult dataset and FRS benefits dataset by their common variables (sernum, benunit and person)

In this case, we choose to keep the numerical version of categorical variables

```
library(foreign)    ### Library necessary to open SPSS files
library(dplyr)      ### Advanced data manipulation functions
library(Hmisc)      ### Extra statistical functions

setwd("/home/piet/Dropbox/work/UKDS/FRS workshop/UKDA-8948-spss/spss/spss25/")
a21<-read.spss("adult.sav",to.data.frame = T,
use.value.labels = F )%>%
  select(SERNUM,BENUNIT,PERSON,GROSS4)
b21<-read.spss("benefits.sav",to.data.frame = T, use.value.labels = F)%>%
  select(SERNUM,BENUNIT,PERSON, BENEFIT,BENAMT)
both1<-merge(a21,b21,by=c("SERNUM","BENUNIT","PERSON"),all.x=F,all.y=F)

names(both1)<-tolower(names(both1)) ### Changes names to lowercase
```

1.2 Benefit frequencies

Gives the grossed frequency, percentage, cumulative frequency and cumulative percentage of each benefit

```
tmptab<-xtabs(gross4~benefit,both1,drop.unused.levels=T)
cbind(
"Freq"= tmptab, "Pct"=round(100*prop.table(tmptab),1),
"Cum. freq."=cumsum(tmptab), "Cum. pct"= round(cumsum(100*prop.table(tmptab)),2)
)
```

##		Freq	Pct	Cum. freq.	Cum. pct
## 1		682377	1.4	682377	1.40
## 2		605697	1.2	1288074	2.65
## 3		6207126	12.8	7495200	15.41
## 4		1006961	2.1	8502161	17.48
## 5		11157837	22.9	19659998	40.43
## 6		53610	0.1	19713608	40.54
## 8		81338	0.2	19794946	40.70
## 9		142150	0.3	19937096	41.00
## 10		8515	0.0	19945611	41.01
## 12		813335	1.7	20758946	42.69
## 13		736470	1.5	21495416	44.20
## 14		252917	0.5	21748333	44.72
## 15		123434	0.3	21871767	44.97
## 16		1425025	2.9	23296792	47.90
## 19		244486	0.5	23541278	48.41
## 21		16183	0.0	23557461	48.44
## 22		18430	0.0	23575891	48.48
## 24		10906	0.0	23586797	48.50
## 32		2449	0.0	23589246	48.51
## 33		22348	0.0	23611594	48.55
## 34		7461	0.0	23619055	48.57

```
## 37      2297  0.0   23621352   48.57
## 61      1300  0.0   23622652   48.57
## 62 11361568 23.4   34984220   71.94
## 65      32452  0.1   35016672   72.00
## 66      57514  0.1   35074186   72.12
## 69      58364  0.1   35132550   72.24
## 70     138664  0.3   35271214   72.53
## 78      6089  0.0   35277303   72.54
## 81      4545  0.0   35281848   72.55
## 82     13705  0.0   35295553   72.58
## 83      1306  0.0   35296859   72.58
## 90     825829  1.7   36122688   74.28
## 91    1431594  2.9   37554282   77.22
## 94    2532242  5.2   40086524   82.43
## 95    2774958  5.7   42861482   88.14
## 96    2175755  4.5   45037237   92.61
## 97    1767003  3.6   46804240   96.24
## 98    303494  0.6   47107734   96.87
## 99      9577  0.0   47117311   96.89
## 110   678344  1.4   47795655   98.28
## 111   831922  1.7   48627577   99.99
## 115     3754  0.0   48631331  100.00
```

1.3 Distribution of Universal Credit, State Pension , Personal Independence Payment (Daily Living), Personal Independence Payment (PIP) Mobility

Examines the distribution of grossed Universal Credit benefit, grossed State Pension benefit amounts, grossed Personal Independence Payment (Daily Living benefit), grossed Personal Independence Payment (PIP) Mobility benefit amounts, including an assessment of normality and discovery of outliers

`xtabs()` produces the tables; `prop.table()` computes the proportions, `round(...,1)` rounds the result to one decimal. R is programmed with parsimony in mind: we only get the output we asked for. This is why we need to specify each indicator we are interested in. We can then gather them all in a single table with `cbind()`. We select the subset of respondents on UC with the `filter()` command. The pipe `%>%` operators allow us to combine several command one after the other.

```
#knitr::kable(
round(t(both11%>%
  filter(benefit==95 | benefit==5 | benefit==96 |
    benefit==97)%>%group_by(benefit)%>%
  summarise(
    "Observations"=length(benamt),
    "Mean"=wtd.mean(benamt,gross4),
    "Median"=wtd.quantile(benamt,gross4,probs=.5),
    "Std Deviation"=sqrt(wtd.var(benamt,gross4)),
    "Variance"=wtd.var(benamt,gross4),
    "Mode"=as.numeric(names(sort(-table(benamt)))[1]),
    "Range"=max(benamt)-min(benamt),
    "IQR"=wtd.quantile(benamt,gross4,probs=.75)-wtd.quantile(benamt,gross4,probs=.25),
    "Skewness"=Weighted.Desc.Stat::w.skewness(benamt,gross4),
    "Kurtosis"=Weighted.Desc.Stat::w.kurtosis(benamt,gross4)
  )),2) #)
```

```
##          [,1]      [,2]      [,3]      [,4]
```

```
## benefit      5.00    95.00  96.00  97.00
## Observations 5573.00 642.00 678.00 566.00
## Mean        157.73  180.86  72.79  46.75
## Median      159.25  169.07  59.70  62.25
## Std Deviation 48.85  108.03  15.25  18.94
## Variance     2386.61 11670.59 232.61 358.81
## Mode         175.00    0.00  59.70  62.25
## Range        475.00  758.23  65.55  38.65
## IQR          46.15  160.21  29.45  38.65
## Skewness      0.30    0.49 -0.04 -0.40
## Kurtosis      2.10    0.39 -1.26 -1.84
```

Weighted quantiles

```
#knitr::kable(
### Computing the results first, and storing them in tmp
tmp<-both1%>%filter(benefit==95| benefit==5 | benefit==96 | benefit==97) %>%
  group_by(benefit)%>%
  summarise(
"ben nr"=wtd.quantile(benamnt,gross4,probs=c(.01,.05,.1,.25,.5,.75,.9,.95,.99))
) #)

### tmp is a long table with each set of nine quantiles
### added one after the other (36 rows).
### The code below makes it easier to read

data.frame(
  "Quantile"=c(.01,.05,.1,.25,.5,.75,.9,.95,.99),
  "UC"=      tmp[tmp$benefit==95,2],
  "Pension"= tmp[tmp$benefit==5,2],
  "PIP (DL)"=tmp[tmp$benefit==96,2],
  "PIP (M)"= tmp[tmp$benefit==97,2])
```

```
## Quantile    ben.nr ben.nr.1 ben.nr.2 ben.nr.3
## 1      0.01    0.00000    36.49    59.70    23.60
## 2      0.05    0.00000    80.00    59.70    23.60
## 3      0.10   50.43912    87.00    59.70    23.60
## 4      0.25   94.33085   134.81    59.70    23.60
## 5      0.50  169.07014   159.25    59.70    62.25
## 6      0.75  254.54071   180.96    89.15    62.25
## 7      0.90  333.69863   211.25    89.15    62.25
## 8      0.95  362.00000   238.58    89.15    62.25
## 9      0.99  432.29622   297.50    89.15    62.25
```

Extreme values

We have chosen to collate the results with `cbind()` in this case, but each part can be run separately if needed.

```
#knitr::kable(
round(
cbind(
  both1%>%filter(benefit==95 | benefit==5 | benefit==96 | benefit==97)%>%
    select(sernum, person, benamt, benefit)%>%
    group_by(benefit)%>%slice_min(order_by = benamt,n=5,with_ties = FALSE),
  both1%>%filter(benefit==95 | benefit==5 | benefit==96 | benefit==97)%>%
    select(sernum, person, benamt, benefit)%>%
    group_by(benefit)%>%slice_max(order_by = benamt,n=5,with_ties = FALSE)
```

```
)
,1)#)

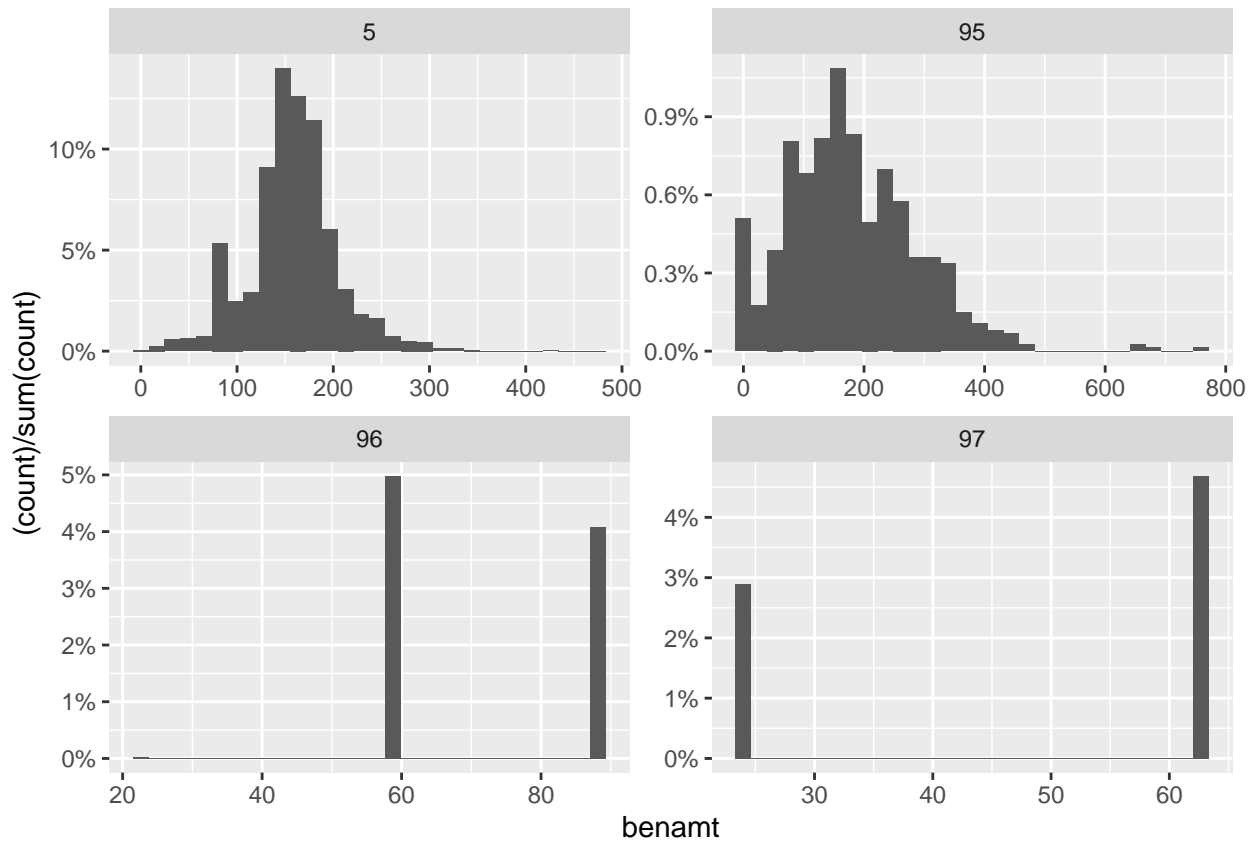
## # A tibble: 20 x 8
##   sernum...1 person...2 benamt...3 benefit...4 sernum~1 perso~2 benam~3 benef~4
##   <dbl>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>
## 1      7709          1          0          5      1200          1      475          5
## 2      1890          2          1          5      5630          1      425          5
## 3      3811          1          1          5       204          1      425.          5
## 4      2589          1         6.2          5      7363          1      355.          5
## 5      3995          1         8.6          5      4913          2      352.          5
## 6      1038          1          0         95      2992          1      758.         95
## 7      1560          1          0         95      2229          1      679.         95
## 8      1607          1          0         95      4292          1      646.         95
## 9      1808          2          0         95      7751          1      645.         95
## 10     1896          1          0         95      9571          1      480.         95
## 11     3237          1         23.6         96      1015          2       89.2         96
## 12     5924          1         23.6         96      1041          1       89.2         96
## 13    10007          2         59.7         96      1065          1       89.2         96
## 14     1006          1         59.7         96      1163          1       89.2         96
## 15     1015          1         59.7         96      1172          2       89.2         96
## 16     1041          1         23.6         97     10007          2       62.2         97
## 17     1120          1         23.6         97      1006          1       62.2         97
## 18     1172          2         23.6         97      1015          2       62.2         97
## 19     1191          1         23.6         97       105          1       62.2         97
## 20     1199          1         23.6         97      1065          1       62.2         97
## # ... with abbreviated variable names 1: sernum...5, 2: person...6,
## #   3: benamt...7, 4: benefit...8
```

1.4 Histograms

Universal credit

The line below creates the histogram.

```
ggplot(both1%>%
  filter(benefit==95 | benefit==5 | benefit==96 | benefit==97),
  aes(x=benamt))+ ### Base plot specification
geom_histogram(aes(y = (..count..)/sum(..count..)))+ ### Converts y axis to percent
scale_y_continuous(labels = scales::percent)+ ### Prints y axis as percent
facet_wrap(~benefit,scale="free") ### Each histogram has its own plot
```



1.5 Distribution of Universal Credit ex. £0

Note There is no need to create a new dataset, we can keep on working with both1

Examines the distribution of grossed Universal Credit benefit amounts excluding £0 observations, including an assessment of normality and discovery of outliers

In the syntax below we exclude cases with £0 benefits by adding *benamt>0* to the *filter()* command.

```
round(t(both1%>%filter(benefit==95 & benamt>0)%>%summarise(
  "Observations"=length(benamt),
  "Mean"=wtd.mean(benamt,gross4),
  "Median"=wtd.quantile(benamt,gross4,probs=.5),
  "Std Deviation"=sqrt(wtd.var(benamt,gross4)),
  "Variance"=wtd.var(benamt,gross4),
  "Mode"=as.numeric(names(sort(-table(benamt)))[1]),
  "Range"=max(benamt)-min(benamt),
  "IQR"=wtd.quantile(benamt,gross4,probs=.75)-wtd.quantile(benamt,gross4,probs=.25),
  "Skewness"=Weighted.Desc.Stat::w.skewness(benamt,gross4),
  "Kurtosis"=Weighted.Desc.Stat::w.kurtosis(benamt,gross4)
)),2)
```

```
##           [,1]
## Observations 611.00
## Mean         191.99
## Median       173.02
## Std Deviation 101.25
## Variance     10251.55
## Mode         94.33
```

```
## Range          757.32
## IQR            140.74
## Skewness       0.65
## Kurtosis       0.68
```

Weighted quantiles

```
both1%>%filter(benefit==95 & benamt>0)%>%select(benamt,gross4)%>%summarise(
  "Quantiles"=
  round(wtd.quantile(benamt,gross4,probs=c(.01,.05,.1,.25,.5,.75,.9,.95,.99)),2)
)
```

```
##   Quantiles
## 1      21.60
## 2      51.99
## 3      74.04
## 4     116.40
## 5     173.02
## 6     257.15
## 7     335.31
## 8     362.81
## 9     432.30
```

Extreme values

```
cbind(
  both1%>%filter(benefit==95 & benamt>0)%>%select(sernum,benunit,person,benamt)%>%
  slice_min(order_by = benamt,n=5,with_ties = FALSE),
  both1%>%filter(benefit==95 & benamt>0)%>%select(sernum,benunit,person,benamt)%>%
  slice_max(order_by = benamt,n=5,with_ties = FALSE)
)
```

```
##   sernum benunit person   benamt sernum benunit person   benamt
## 1   4183      2      3 0.9182466  2992      1      1 758.2346
## 2   7667      4      4 1.8802192  2229      1      1 679.3828
## 3   6337      1      1 3.3784110  4292      1      1 645.5872
## 4   6814      1      1 3.5349041  7751      1      1 645.1706
## 5   5498      1      1 3.6499726  9571      1      1 479.5600
```

Histogram of Universal Credit (without 0s)

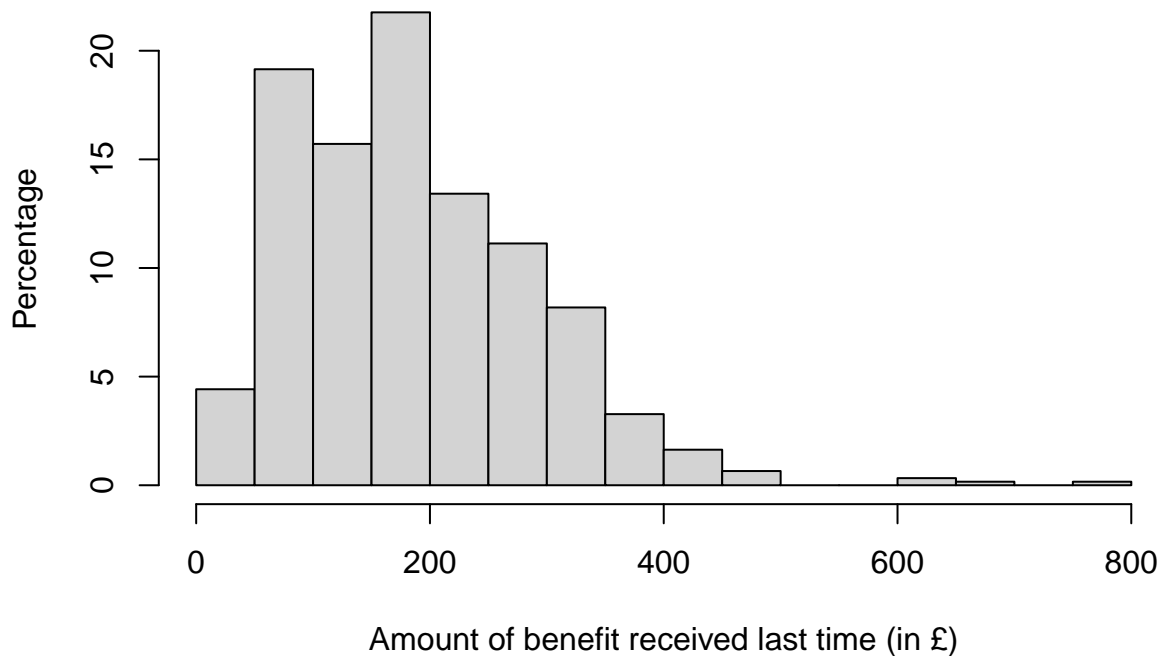
The syntax below also select a subset of the data, with square brackets as an alternative to the *filter()* command used above. We also use an alternative command to create the histogram.

```
z<-hist(both1$benamt[both1$benefit==95 & both1$benamt>0],freq=F)
```

```
z$density <- z$counts/sum(z$counts) * 100 ### Converting y axis into percentages
```

```
plot(z, freq = FALSE,
     main= '% of amount of benefit received for those on PIP Mobility',
     xlab="Amount of benefit received last time (in £)",
     ylab="Percentage"
)
```

% of amount of benefit received for those on PIP Mobility



1.6 UC and tenure

Create uc_tenure2

Create a new dataset called 'uc_tenure2' from merging the 'uc_tenure1' (ie both1) dataset and FRS household dataset by their common variable (sernum)

```
setwd("/home/piet/Dropbox/work/UKDS/FRS workshop/UKDA-8948-spss/spss/spss25" )
h21<-read.spss("househol.sav",to.data.frame = T, use.value.labels = T,
add.undeclared.levels="append")%>%select(SERNUM,TENURE)
names(h21)<-tolower(names(h21))
uc_tenure2<-merge(both1,h21,by=c("sernum"),all.x=F,all.y=F)
levels(uc_tenure2$tenure)[2]<-"Bought with mortgage" ### Shorten a category, for nicer display
```

Frequency of Tenure

Gives the grossed frequency, percentage, cumulative frequency and cumulative percentage of each category for the variable 'tenure' for those in receipt of Universal Credit

```
tmptab<-xtabs(gross4~tenure,uc_tenure2,subset=benefit==95,drop.unused.levels=T)
cbind(
"Freq"= tmptab, "Pct"=round(100*prop.table(tmptab),1),
"Cum. freq."=cumsum(tmptab), "Cum. pct"= round(cumsum(100*prop.table(tmptab)),2)
)
```

##	Freq	Pct	Cum. freq.	Cum. pct
## Owns it outright	243349	8.8	243349	8.77
## Bought with mortgage	282334	10.2	525683	18.94
## Part own, part rent	15214	0.5	540897	19.49
## Rents	2217836	79.9	2758733	99.42
## Rent-free	16225	0.6	2774958	100.00

2 Benefit Comparison – Year: 2019-20

Create a new dataset Both3 (we could re-use both1) from merging the FRS adult dataset and FRS benefits dataset by their common variables (sernum, benunit and person)

```
setwd("/home/piet/Dropbox/work/UKDS/FRS workshop/UKDA-8802-spss/spss/spss25/")
a19<-read.spss("adult.sav",to.data.frame = T,use.value.labels = F )%>%
select(SERNUM,BENUNIT,PERSON,GROSS4)
b19<-read.spss("benefits.sav",to.data.frame = T,use.value.labels = F)%>%
select(SERNUM,BENUNIT,PERSON, BENEFIT,BENAMT)
both3<-merge(a19,b19,by=c("SERNUM","BENUNIT","PERSON"),all.x=F,all.y=F)

names(both3)<-tolower(names(both3))
```

2.1 Benefit frequencies

Gives the grossed frequency, percentage, cumulative frequency and cumulative percentage of each benefit

```
tmptab<-xtabs(gross4~benefit,both3,drop.unused.levels=T)
cbind(
"Freq"= tmptab,
"Pct"=round(100*prop.table(tmptab),1),
"Cum. freq."=cumsum(tmptab),
"Cum. pct"= round(cumsum(100*prop.table(tmptab)),2)
)
```

##		Freq	Pct	Cum. freq.	Cum. pct
## 1	1003954	2.1	1003954	2.08	
## 2	816758	1.7	1820712	3.77	
## 3	6260230	12.9	8080942	16.71	
## 4	1128441	2.3	9209383	19.05	
## 5	11302238	23.4	20511621	42.42	
## 6	57256	0.1	20568877	42.54	
## 8	58179	0.1	20627056	42.66	
## 9	114847	0.2	20741903	42.90	
## 10	15110	0.0	20757013	42.93	
## 12	814195	1.7	21571208	44.62	
## 13	662945	1.4	22234153	45.99	
## 14	241009	0.5	22475162	46.49	
## 15	152041	0.3	22627203	46.80	
## 16	1244643	2.6	23871846	49.37	
## 19	378065	0.8	24249911	50.16	
## 21	13961	0.0	24263872	50.19	
## 22	28481	0.1	24292353	50.24	
## 24	10723	0.0	24303076	50.27	
## 31	10803	0.0	24313879	50.29	
## 32	1771	0.0	24315650	50.29	
## 33	60499	0.1	24376149	50.42	
## 34	7752	0.0	24383901	50.43	
## 35	12546	0.0	24396447	50.46	
## 36	9893	0.0	24406340	50.48	
## 37	5510	0.0	24411850	50.49	
## 61	6878	0.0	24418728	50.51	
## 62	11479871	23.7	35898599	74.25	

```
## 65      44592 0.1  35943191  74.34
## 66      67486 0.1  36010677  74.48
## 69      92289 0.2  36102966  74.67
## 70     130467 0.3  36233433  74.94
## 78      10190 0.0  36243623  74.96
## 81       5167 0.0  36248790  74.97
## 82      14097 0.0  36262887  75.00
## 83       5313 0.0  36268200  75.01
## 90     1035774 2.1  37303974  77.16
## 91     1870822 3.9  39174796  81.03
## 94     2860448 5.9  42035244  86.94
## 95     1376924 2.8  43412168  89.79
## 96     1577237 3.3  44989405  93.05
## 97     1176666 2.4  46166071  95.49
## 98       50744 0.1  46216815  95.59
## 99      349210 0.7  46566025  96.31
## 102      45448 0.1  46611473  96.41
## 103     136546 0.3  46748019  96.69
## 104      54263 0.1  46802282  96.80
## 105     28931 0.1  46831213  96.86
## 106     42498 0.1  46873711  96.95
## 107     18598 0.0  46892309  96.99
## 108     36122 0.1  46928431  97.06
## 109     95016 0.2  47023447  97.26
## 110     77910 0.2  47101357  97.42
## 111     381552 0.8  47482909  98.21
## 112     191637 0.4  47674546  98.61
## 113     618294 1.3  48292840  99.89
## 114      55232 0.1  48348072 100.00
```

2.2 Distribution of Universal Credit, State Pension , Personal Independence Payment (Daily Living), Personal Independence Payment (PIP) Mobility

Examines the distribution of grossed Universal Credit benefit, State Pension benefit amounts, Personal Independence Payment (Daily Living benefit), Personal Independence Payment (PIP) Mobility benefit amounts, including an assessment of normality and discovery of outliers

```
round(t(both3%>%filter(benefit==95 | benefit==5 | benefit==96 | benefit==97)%>%
  group_by(benefit)%>%summarise(
    "Observations"=length(benamt),
    "Mean"=wtd.mean(benamt,gross4),
    "Median"=wtd.quantile(benamt,gross4,probs=.5),
    "Std Deviation"=sqrt(wtd.var(benamt,gross4)),
    "Variance"=wtd.var(benamt,gross4),
    "Mode"=as.numeric(names(sort(-table(benamt)))[1]),
    "Range"=max(benamt)-min(benamt),
    "IQR"=wtd.quantile(benamt,gross4,probs=.75)-wtd.quantile(benamt,gross4,probs=.25),
    "Skewness"=Weighted.Desc.Stat::w.skewness(benamt,gross4),
    "Kurtosis"=Weighted.Desc.Stat::w.kurtosis(benamt,gross4)
  )),2) #)
```

```
##           [,1]    [,2]    [,3]    [,4]
## benefit      5.00    95.00    96.00    97.00
## Observations 9055.00  919.00 1204.00 905.00
## Mean        150.15  162.92   71.43  44.93
```

```
## Median      150.38  151.03   58.70  61.20
## Std Deviation  46.76   95.73   14.37  18.84
## Variance    2186.20 9164.64  206.47 355.08
## Mode        150.00   0.00   58.70  61.20
## Range       433.76  507.00   28.95  96.70
## IQR         45.65  148.92   28.95  38.00
## Skewness     0.14   0.50    0.24 -0.27
## Kurtosis     1.33   0.00   -1.94 -1.86
```

Weighted quantiles

```
#knitr::kable(
### Computing the results first, and storing them in tmp
tmp<-both3%>%filter(benefit==95| benefit==5 | benefit==96 | benefit==97) %>%
  group_by(benefit)%>% summarise(
    "ben#"=wtd.quantile(benam, gross4, probs=c(.01,.05,.1,.25,.5,.75,.9,.95,.99))
  ) #)
```

```
### tmp is a long table with each set of nine quantiles
### added one after the other (36 rows).
### The code below makes it easier to read
```

```
data.frame(
  "Quantile"=c(.01,.05,.1,.25,.5,.75,.9,.95,.99),
  "UC"=      tmp[tmp$benefit==95,2],
  "Pension"= tmp[tmp$benefit==5,2],
  "PIP (DL)"=tmp[tmp$benefit==96,2],
  "PIP (M)"= tmp[tmp$benefit==97,2])
```

```
## Quantile      ben.    ben..1 ben..2 ben..3
## 1      0.01    0.00000  34.0500  58.70   23.2
## 2      0.05   28.76197  75.0000  58.70   23.2
## 3      0.10   46.02740  80.8000  58.70   23.2
## 4      0.25   79.17344 129.2000  58.70   23.2
## 5      0.50  151.03016 150.3775  58.70   61.2
## 6      0.75  228.09213 174.8500  87.65   61.2
## 7      0.90  287.06885 204.4300  87.65   61.2
## 8      0.95  326.41803 227.6700  87.65   61.2
## 9      0.99  428.95082 279.7500  87.65   61.2
```

Extreme values

We have chosen to collate the results with `cbind()` in this case, but each part can be run separately if needed.

```
#knitr::kable(
round(
cbind(
  both3%>%filter(benefit==95 | benefit==5 | benefit==96 | benefit==97)%>%
    select(sernum, person, benamt, benefit)%>%
    group_by(benefit)%>%slice_min(order_by = benamt, n=5, with_ties = FALSE),
  both3%>%filter(benefit==95 | benefit==5 | benefit==96 | benefit==97)%>%
    select(sernum, person, benamt, benefit)%>%
    group_by(benefit)%>%slice_max(order_by = benamt, n=5, with_ties = FALSE)
),
1) #)
```

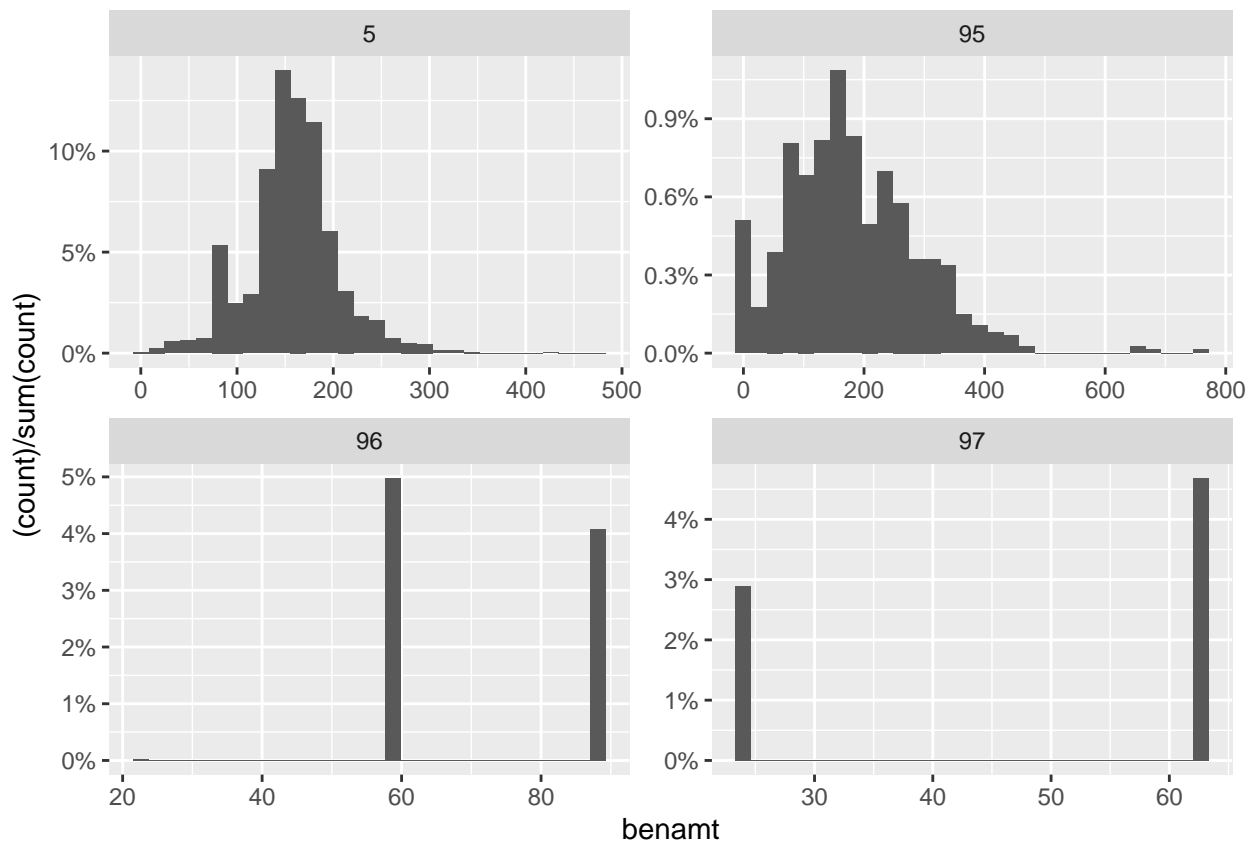
```
## # A tibble: 20 x 8
```

```
##      sernum...1 person...2 benamt...3 benefit...4 sernum~1 perso~2 benam~3 benef~4
##      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>
##  1      9257          3          0          5      2727          2      434.          5
##  2      3302          2          0.1        5     13813          1      388          5
##  3     10263          2          0.2        5     17339          1     380.          5
##  4     12153          1          0.2        5     17521          1     372.          5
##  5     14864          1          0.8        5      2544          1     354          5
##  6     10810          5          0          95     14585          2     507          95
##  7     11967          1          0          95     14561          1     483          95
##  8     12635          2          0          95     10708          1     471          95
##  9     12884          1          0          95     10705          1     470.          95
## 10     15995          1          0          95      7993          1     452.          95
## 11     10018          1        58.7         96     10019          2      87.7         96
## 12     10023          1        58.7         96      1002          1      87.7         96
## 13     10050          1        58.7         96     10031          1      87.7         96
## 14     10081          1        58.7         96     10035          1      87.7         96
## 15     10158          1        58.7         96     10096          1      87.7         96
## 16     10018          1        23.2         97     11765          1     120.          97
## 17     10019          2        23.2         97     10035          1      61.2         97
## 18      1002          1        23.2         97     10096          1      61.2         97
## 19     10031          1        23.2         97     10098          1      61.2         97
## 20     10050          1        23.2         97     10123          1      61.2         97
## # ... with abbreviated variable names 1: sernum...5, 2: person...6,
## #   3: benamt...7, 4: benefit...8
```

2.3 Histograms

Universal credit

```
ggplot(both1%>%
  filter(benefit==95 | benefit==5 | benefit==96 | benefit==97),
  aes(x=benamt))+ ### Base plot specification
geom_histogram(aes(y = (..count..)/sum(..count..)))+ ### Converts y axis to percent
scale_y_continuous(labels = scales::percent)+ ### Prints y axis as percent
facet_wrap(~benefit,scale="free") ### Each histogram has its own plot
```



2.4 Create Both2 – Distribution of Universal Credit ex. £0

(Note There is no need for this step in R, we can keep on working with both3)

Examines the distribution of grossed Universal Credit benefit amounts excluding £0 observations, including an assessment of normality and discovery of outliers

```
round(t(both3%>%filter(benefit==95 & benamt>0)%>%summarise(
  "Observations"=length(benamt),
  "Mean"=wtd.mean(benamt,gross4),
  "Median"=wtd.quantile(benamt,gross4,probs=.5),
  "Std Deviation"=sqrt(wtd.var(benamt,gross4)),
  "Variance"=wtd.var(benamt,gross4),
  "Mode"=as.numeric(names(sort(-table(benamt)))[1]),
  "Range"=max(benamt)-min(benamt),
  "IQR"=wtd.quantile(benamt,gross4,probs=.75)-wtd.quantile(benamt,gross4,probs=.25),
  "Skewness"=Weighted.Desc.Stat::w.skewness(benamt,gross4),
  "Kurtosis"=Weighted.Desc.Stat::w.kurtosis(benamt,gross4)
)),2)
```

```
##           [,1]
## Observations 902.00
## Mean         166.64
## Median        155.15
## Std Deviation  93.56
## Variance      8754.38
## Mode          72.94
## Range        506.71
```

```
## IQR          138.39
## Skewness     0.55
## Kurtosis     0.05
```

Weighted quantiles

```
both3%>%filter(benefit==95 & benamt>0)%>%select(benamt,gross4)%>%summarise(
  "Quantiles"=
  round(wtd.quantile(benamt,gross4,probs=c(.01,.05,.1,.25,.5,.75,.9,.95,.99)),2)
)
```

```
##   Quantiles
## 1      6.56
## 2     38.84
## 3     52.93
## 4     92.05
## 5    155.15
## 6    230.45
## 7    288.46
## 8    326.48
## 9    428.95
```

Extreme values

```
cbind(
  both3%>%filter(benefit==95 & benamt>0)%>%select(sernum,benunit,person,benamt)%>%
  slice_min(order_by = benamt,n=5,with_ties = FALSE),
  both3%>%filter(benefit==95 & benamt>0)%>%select(sernum,benunit,person,benamt)%>%
  slice_max(order_by = benamt,n=5,with_ties = FALSE)
)
```

```
##   sernum benunit person   benamt sernum benunit person   benamt
## 1   5504      1      1 0.2853699 14585      1      2 507.0000
## 2   2748      1      1 0.3544110 14561      1      1 482.9564
## 3  14014      1      1 0.4268852 10708      1      1 470.9692
## 4  16360      2      2 1.1429508 10705      1      1 470.5400
## 5   5746      1      2 1.4504918  7993      1      1 451.7502
```

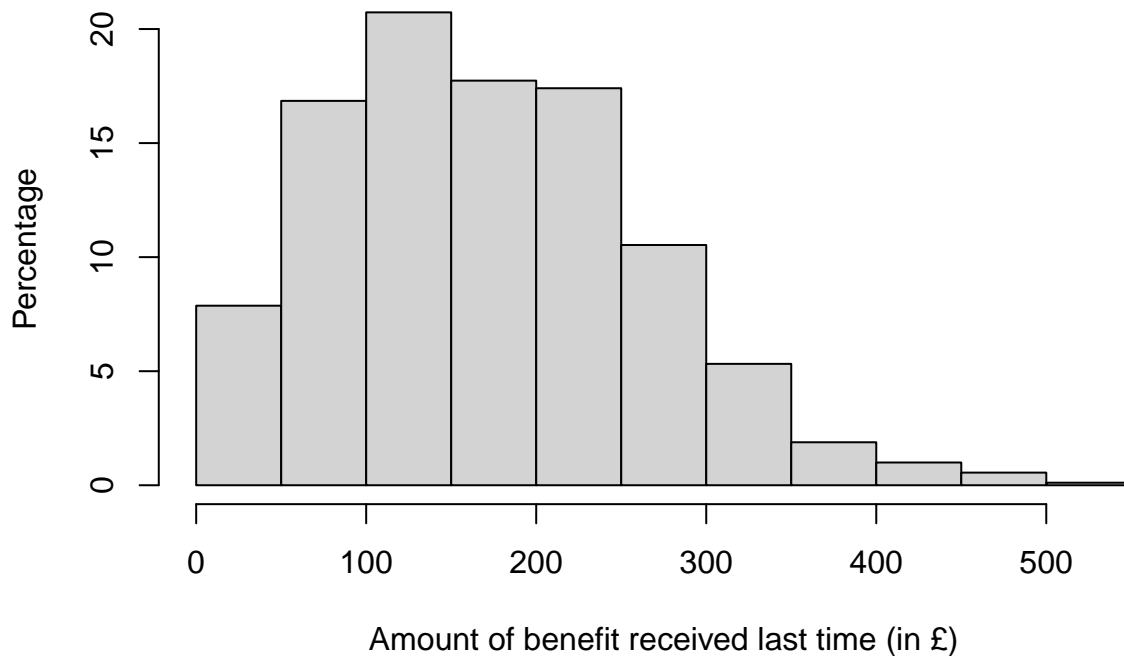
**** Histogram of Universal Credit (without 0s)****

```
z<-hist(both3$benamt[both3$benefit==95 & both3$benamt>0],freq=F)
```

```
z$density <- z$counts/sum(z$counts) * 100 ### Converting y axis into percentages
```

```
plot(z, freq = FALSE,
     main= '% of amount of benefit received for those on PIP Mobility',
     xlab="Amount of benefit received last time (in £)",
     ylab="Percentage"
)
```

% of amount of benefit received for those on PIP Mobility



2.5 UC and tenure

Create tenure4 Create a new dataset called 'uc_tenure4' from merging the 'uc_tenure1'(ie both3) dataset and FRS household dataset by their common variable (sernum)**

```
setwd("/home/piet/Dropbox/work/UKDS/FRS workshop/UKDA-8948-spss/spss/spss25" )
h21<-read.spss("househol.sav",to.data.frame = T, use.value.labels = T,
add.undeclared.levels="append")%>%select(SERNUM,TENURE)
names(h21)<-tolower(names(h21))
uc_tenure4<-merge(both3,h21,by=c("sernum"),all.x=F,all.y=F)
levels(uc_tenure4$tenure)[2]<-"Bought with mortgage"
```

Frequency of Tenure

Gives the grossed frequency, percentage, cumulative frequency and cumulative percentage of each category for the variable 'tenure' for those in receipt of Universal Credit

```
tmptab<-xtabs(gross4~tenure,uc_tenure4,subset=benefit==95,drop.unused.levels=T)
cbind(
  "Freq"= tmptab,
  "Pct"=round(100*prop.table(tmptab),1),
  "Cum. freq."=cumsum(tmptab),
  "Cum. pct"= round(cumsum(100*prop.table(tmptab)),2)
)
```

##		Freq	Pct	Cum. freq.	Cum. pct
##	Owns it outright	324217	46.0	324217	45.99
##	Bought with mortgage	176384	25.0	500601	71.02
##	Part own, part rent	1142	0.2	501743	71.18
##	Rents	198607	28.2	700350	99.35
##	Rent-free	4561	0.6	704911	100.00

3 Statistically significant differences between 2019-20 and 2020-21

3.1 Create Combine_years

Create a new appended dataset called 'combine_years' that contains the 'ad_ben_1920' dataset and 'ad_ben_2021' dataset (ie Both1 and Both3) A new variable called 'year' is created that assigns '1920' to the case if the year is 2019-20 and '2021' if the year is 2020-21

```
combine_years<-rbind(  
  cbind(both1,year=2021),  
  cbind(both3,year=1920)  
)
```

/Provides grossed analysis of variance and Kruskal-Wallis for Universal Credit benefit amounts by year/

3.2 ANOVA UC amount by year

```
summary(lm(benamt~as.factor(year),data=combine_years%>%filter(benefit==95),weights=gross4))
```

```
##  
## Call:  
## lm(formula = benamt ~ as.factor(year), data = combine_years %>%  
##   filter(benefit == 95), weights = gross4)  
##  
## Weighted Residuals:  
##      Min      1Q   Median      3Q      Max   
## -18448.5 -3073.7  -438.3   2896.9  26313.4   
##  
## Coefficients:  
##              Estimate Std. Error t value Pr(>|t|)      
## (Intercept)      162.921      4.579   35.582 < 2e-16 ***  
## as.factor(year)2021    17.938      5.601    3.203  0.00139 **  
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 5373 on 1559 degrees of freedom  
## Multiple R-squared:  0.006537, Adjusted R-squared:  0.0059  
## F-statistic: 10.26 on 1 and 1559 DF, p-value: 0.001388  
  
### We need to use the Survey package in order to access the weighted rank test functions  
### (Wilcoxon/Kruskal-Wallis)  
library(survey)  
  
### We also need to declare the survey design, in this case we will only specify the  
### grossing weights  
design <- svydesign(ids = ~0, data=combine_years%>%filter(benefit==95),  
weights=combine_years%>%filter(benefit==95)%>%select(gross4))  
  
svyranktest(benamt ~ as.factor(year), design = design)  
  
##  
## Design-based KruskalWallis test  
##  
## data:  benamt ~ as.factor(year)  
## t = 2.7011, df = 1559, p-value = 0.006987  
## alternative hypothesis: true difference in mean rank score is not equal to 0
```



```
## sample estimates:
## difference in mean rank score
## 0.04835153
```

3.3 ANOVA State pension benefit amount by year

/Provides grossed analysis of variance and Kruskal-Wallis Rank Sums for State Pension benefit amounts by year/

```
summary(lm(benamt~as.factor(year),data=combine_years%>%filter(benefit==5),weights=gross4))
```

```
##
## Call:
## lm(formula = benamt ~ as.factor(year), data = combine_years %>%
##   filter(benefit == 5), weights = gross4)
##
## Weighted Residuals:
##      Min       1Q   Median       3Q      Max
## -14600.3   -776.5    -4.8    818.9   30003.1
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    150.1536     0.5573  269.438  <2e-16 ***
## as.factor(year)2021    7.5813     0.7907   9.589  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1874 on 14626 degrees of freedom
## Multiple R-squared:  0.006247, Adjusted R-squared:  0.006179
## F-statistic: 91.94 on 1 and 14626 DF, p-value: < 2.2e-16
```

```
design <- svydesign(ids = ~0, data=combine_years%>%filter(benefit==5),
weights=combine_years%>%filter(benefit==5)%>%select(gross4))
```

```
svyranktest(benamt ~ as.factor(year), design = design)
```

```
##
## Design-based KruskalWallis test
##
## data:  benamt ~ as.factor(year)
## t = 8.5945, df = 14626, p-value < 2.2e-16
## alternative hypothesis: true difference in mean rank score is not equal to 0
## sample estimates:
## difference in mean rank score
## 0.05344663
```

3.4 ANOVA PIP Daily Living

/Provides grossed analysis of variance and Kruskal-Wallis Rank Sums for Personal Independence Payment (PIP) Daily Living benefit amounts by year/

```
summary(lm(benamt~as.factor(year),data=combine_years%>%filter(benefit==96),weights=gross4))
```

```
##
## Call:
## lm(formula = benamt ~ as.factor(year), data = combine_years %>%
```

```
##      filter(benefit == 96), weights = gross4)
##
## Weighted Residuals:
##      Min      1Q  Median      3Q      Max
## -5017.9  -490.6  -319.5   590.1  1871.6
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      71.4256    0.5296 134.861  <2e-16 ***
## as.factor(year)2021  1.3692    0.6956   1.968   0.0492 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 665.1 on 1880 degrees of freedom
## Multiple R-squared:  0.002057, Adjusted R-squared:  0.001526
## F-statistic: 3.875 on 1 and 1880 DF, p-value: 0.04917
design <- svydesign(ids = ~0, data=combine_years%>%filter(benefit==96),
weights=combine_years%>%filter(benefit==96)%>%select(gross4))

svyranktest(benamt ~ as.factor(year), design = design)

##
## Design-based KruskalWallis test
##
## data:  benamt ~ as.factor(year)
## t = 16.83, df = 1880, p-value < 2.2e-16
## alternative hypothesis: true difference in mean rank score is not equal to 0
## sample estimates:
## difference in mean rank score
##              0.2546178
```

3.5 ANOVA PIP mobility

/Provides grossed analysis of variance and Kruskal-Wallis Rank Sums for Personal Independence Payment (PIP) Mobility benefit amounts by year/

```
summary(lm(benamt~as.factor(year),data=combine_years%>%filter(benefit==97),weights=gross4))

##
## Call:
## lm(formula = benamt ~ as.factor(year), data = combine_years %>%
##      filter(benefit == 97), weights = gross4)
##
## Weighted Residuals:
##      Min      1Q  Median      3Q      Max
## -2361.6  -790.0   413.2   625.3  1773.5
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      44.9285    0.7801  57.595  <2e-16 ***
## as.factor(year)2021  1.8233    1.0068   1.811   0.0704 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 846.2 on 1469 degrees of freedom
```

```
## Multiple R-squared:  0.002227,   Adjusted R-squared:  0.001548
## F-statistic: 3.279 on 1 and 1469 DF,  p-value: 0.07036

design <- svydesign(ids = ~0, data=combine_years%>%filter(benefit==97),
weights=combine_years%>%filter(benefit==97)%>%select(gross4))

svyranktest(benamt ~ as.factor(year), design = design)

##
## Design-based KruskalWallis test
##
## data:  benamt ~ as.factor(year)
## t = 16.198, df = 1469, p-value < 2.2e-16
## alternative hypothesis: true difference in mean rank score is not equal to 0
## sample estimates:
## difference in mean rank score
## 0.268868
```

4 Wellbeing variables

4.1 Comparison between 2019-20 and 2020-21

Creating a 'combine_years' dataset from the merged individual and households datasets from 2019-20 and 2020-21

```
setwd("/home/piet/Dropbox/work/UKDS/FRS workshop/")
combine_years<-merge(
rbind(
cbind(read.spss("UKDA-8948-spss/spss/spss25/adult.sav",to.data.frame = T,
use.value.labels = F)%>%
select(SERNUM,BENUNIT,PERSON,GROSS4,DISD07,HAPPYWB,ANXIOUS,LIFESAT,MEANING),year=1920),

cbind(read.spss("UKDA-8802-spss/spss/spss25/adult.sav",to.data.frame = T,
use.value.labels = F)%>%
select(SERNUM,BENUNIT,PERSON,GROSS4,DISD07,HAPPYWB,ANXIOUS,LIFESAT,MEANING),year=2021)
),

t<-rbind(
cbind(read.spss("UKDA-8948-spss/spss/spss25/househol.sav",to.data.frame = T,
use.value.labels = F)%>%
select(SERNUM,HHINCBND),year=1920),

cbind(read.spss("UKDA-8802-spss/spss/spss25/househol.sav",to.data.frame = T,
use.value.labels = F)%>%
select(SERNUM,HHINCBND),year=2021)
),

by=c("SERNUM","year"),all.x=F,all.y=F)

names(combine_years)<-tolower(names(combine_years))
```

Gives the grossed frequency, percentage, cumulative frequency and cumulative percentage of each category for each wellbeing variable Happiness 2019-20 and 2020-21

```
tmptab<-xtabs(gross4~happywb+year,combine_years) ### creates the table
```

```
### Computes Frequencies and percentages
```

```
cbind(
"Freq"= tmptab,
"Pct"=round(100*prop.table(tmptab),2)
)
```

##		1920	2021	1920	2021
##	0	354257	410207	0.47	0.54
##	1	304245	241372	0.40	0.32
##	2	578424	636717	0.76	0.84
##	3	1117826	931021	1.47	1.22
##	4	1386922	1290415	1.82	1.69
##	5	3269500	3114612	4.29	4.09
##	6	3617610	3109194	4.75	4.08
##	7	6483098	6175554	8.51	8.11
##	8	8905319	9858331	11.69	12.95
##	9	5467150	6359517	7.18	8.35
##	10	5653133	6882457	7.42	9.04

```
#### Cumulated frequencies and percentages
cbind(
  "Cum. freq."=cumsum(as.data.frame.matrix(tmptab)),
  "Cum. pct"= round(cumsum(as.data.frame.matrix(100*prop.table(tmptab,2))),1)
)
```

```
##      Cum. freq..1920 Cum. freq..2021 Cum. pct.1920 Cum. pct.2021
## 0          354257          410207          1.0          1.1
## 1          658502          651579          1.8          1.7
## 2         1236926         1288296          3.3          3.3
## 3         2354752         2219317          6.3          5.7
## 4         3741674         3509732         10.1          9.0
## 5         7011174         6624344         18.9         17.0
## 6        10628784         9733538         28.6         25.0
## 7        17111882        15909092         46.1         40.8
## 8        26017201        25767423         70.1         66.1
## 9        31484351        32126940         84.8         82.4
## 10       37137484        39009397        100.0        100.0
```

Anxiety 2019-20

```
tmptab<-xtabs(gross4~anxious+year,combine_years) ### creates the table
```

```
#### Computes Frequencies and percents
cbind(
  "Freq"= tmptab,
  "Pct"=round(100*prop.table(tmptab),2)
)
```

```
##      1920      2021 1920 2021
## 0 11311437 13180739 14.86 17.31
## 1  3354055  3680690  4.41  4.83
## 2  5132901  5117912  6.74  6.72
## 3  3184121  3279182  4.18  4.31
## 4  2344513  2302439  3.08  3.02
## 5  3824487  3684135  5.02  4.84
## 6  2184132  2000579  2.87  2.63
## 7  2162332  2176659  2.84  2.86
## 8  2051803  1996550  2.69  2.62
## 9   843754   807702  1.11  1.06
## 10  730299   783908  0.96  1.03
```

```
#### Cumulated frequencies and percentages
cbind(
  "Cum. freq."=cumsum(as.data.frame.matrix(tmptab)),
  "Cum. pct"= round(cumsum(as.data.frame.matrix(100*prop.table(tmptab,2))),1)
)
```

```
##      Cum. freq..1920 Cum. freq..2021 Cum. pct.1920 Cum. pct.2021
## 0          11311437          13180739          30.5          33.8
## 1          14665492          16861429          39.5          43.2
## 2          19798393          21979341          53.3          56.3
## 3          22982514          25258523          61.9          64.7
## 4          25327027          27560962          68.2          70.7
## 5          29151514          31245097          78.5          80.1
## 6          31335646          33245676          84.4          85.2
```

## 7	33497978	35422335	90.2	90.8
## 8	35549781	37418885	95.8	95.9
## 9	36393535	38226587	98.0	98.0
## 10	37123834	39010495	100.0	100.0

Life satisfaction 2019-20

```
tmptab<-xtabs(gross4~lifesat+year,combine_years) ### creates the table
```

```
### Computes Frequencies and percents
cbind(
  "Freq"= tmptab,
  "Pct"=round(100*prop.table(tmptab),2)
)
```

##	1920	2021	1920	2021
## 0	307607	303214	0.40	0.40
## 1	318731	358313	0.42	0.47
## 2	350467	372074	0.46	0.49
## 3	736660	574953	0.97	0.76
## 4	1068417	829821	1.40	1.09
## 5	3147252	2865933	4.13	3.76
## 6	3417905	2851165	4.49	3.74
## 7	7315920	6909701	9.61	9.08
## 8	10890719	12414550	14.30	16.31
## 9	4875114	5991986	6.40	7.87
## 10	4694164	5544646	6.17	7.28

```
#### Cumulated frequencies and percentages
cbind(
  "Cum. freq."=cumsum(as.data.frame.matrix(tmptab)),
  "Cum. pct"= round(cumsum(as.data.frame.matrix(100*prop.table(tmptab,2))),1)
)
```

##	Cum. freq..1920	Cum. freq..2021	Cum. pct.1920	Cum. pct.2021
## 0	307607	303214	0.8	0.8
## 1	626338	661527	1.7	1.7
## 2	976805	1033601	2.6	2.6
## 3	1713465	1608554	4.6	4.1
## 4	2781882	2438375	7.5	6.2
## 5	5929134	5304308	16.0	13.6
## 6	9347039	8155473	25.2	20.9
## 7	16662959	15065174	44.9	38.6
## 8	27553678	27479724	74.2	70.4
## 9	32428792	33471710	87.4	85.8
## 10	37122956	39016356	100.0	100.0

Life Meaningfulness 2019-20 and 2020-21

```
tmptab<-xtabs(gross4~meaning+year,combine_years)
```

```
cbind(
  "Freq"= tmptab,
  "Pct"=round(100*prop.table(tmptab),2)
)
```

##	1920	2021	1920	2021
----	------	------	------	------

```
## 0    192049    233193    0.25    0.31
## 1     98069    114816    0.13    0.15
## 2    286656    293746    0.38    0.39
## 3    458307    505038    0.60    0.66
## 4    797482    631245    1.05    0.83
## 5    2703368   2440702    3.56    3.21
## 6    2583599   2554497    3.40    3.36
## 7    6589179   6468443    8.67    8.51
## 8   10907043  11573508   14.34   15.22
## 9    6039536   6664005    7.94    8.76
## 10   6445424   7458907    8.48    9.81
```

Cumulated frequencies and percentages

```
cbind(
  "Cum. freq."=cumsum(as.data.frame.matrix(tmptab)),
  "Cum. pct"= round(cumsum(as.data.frame.matrix(100*prop.table(tmptab,2))),1)
)
```

```
##      Cum. freq..1920 Cum. freq..2021 Cum. pct.1920 Cum. pct.2021
## 0           192049           233193           0.5           0.6
## 1           290118           348009           0.8           0.9
## 2           576774           641755           1.6           1.6
## 3          1035081          1146793           2.8           2.9
## 4          1832563          1778038           4.9           4.6
## 5          4535931          4218740          12.2          10.8
## 6          7119530          6773237          19.2          17.4
## 7          13708709          13241680          36.9          34.0
## 8          24615752          24815188          66.3          63.7
## 9          30655288          31479193          82.6          80.8
## 10         37100712          38938100         100.0         100.0
```

Examines the distribution of happiness in 2019-20 and 2020-21, including an assessment of normality and discovery of outliers

```
round(t(combine_years%>%filter(!is.na(happywb))%>%group_by(year)%>%summarise(
  "Observations"=length(happywb),
  "Mean"=wtd.mean(happywb,gross4),
  "Median"=wtd.quantile(happywb,gross4,probs=.5),
  "Std Deviation"=sqrt(wtd.var(happywb,gross4)),
  "Variance"=wtd.var(happywb,gross4),
  "Mode"=as.numeric(names(sort(-table(happywb)))[1]),
  "Range"=max(happywb)-min(happywb),
  "IQR"=wtd.quantile(happywb,gross4,probs=.75)-wtd.quantile(happywb,gross4,probs=.25),
  "Skewness"=Weighted.Desc.Stat:w.skewness(happywb,gross4),
  "Kurtosis"=Weighted.Desc.Stat:w.kurtosis(happywb,gross4)
)),2)
```

```
##           [,1]    [,2]
## year      1920.00  2021.00
## Observations 12923.00 26262.00
## Mean          7.29    7.48
## Median         8.00    8.00
## Std Deviation  2.13    2.12
## Variance       4.53    4.49
## Mode          8.00    8.00
## Range         10.00   10.00
```

```
## IQR          3.00    2.00
## Skewness     -0.95   -1.09
## Kurtosis      0.84    1.20
```

Weighted quantiles

```
cbind(
  combine_years%>%filter(year==1920 & !is.na(happywb))%>%select(happywb,gross4)%>%summarise(
    "Quantiles"=
    round(wtd.quantile(happywb,gross4,probs=c(.01,.05,.1,.25,.5,.75,.9,.95,.99)),2)
  ),
  combine_years%>%filter(year==1920 & !is.na(happywb))%>%select(happywb,gross4)%>%summarise(
    "Quantiles"=
    round(wtd.quantile(happywb,gross4,probs=c(.01,.05,.1,.25,.5,.75,.9,.95,.99)),2)
  )
)
```

```
##   Quantiles Quantiles
## 1         1         1
## 2         3         3
## 3         4         4
## 4         6         6
## 5         8         8
## 6         9         9
## 7        10        10
## 8        10        10
## 9        10        10
```

Extreme values

```
cbind(
  combine_years%>%filter(year==1920 & !is.na(happywb))%>%select(happywb)%>%
  slice_min(order_by = happywb,n=5,with_ties = FALSE),
  combine_years%>%filter(year==2021 & !is.na(happywb))%>%select(happywb)%>%
  slice_max(order_by = happywb,n=5,with_ties = FALSE)
)
```

```
##   happywb happywb
## 1       0       10
## 2       0       10
## 3       0       10
## 4       0       10
## 5       0       10
```

Examines the distribution of anxiety in 2019-20 and 2020-21, including an assessment of normality and discovery of outliers

```
round(t(combine_years%>%filter(!is.na(anxious))%>%group_by(year)%>%summarise(
  "Observations"=length(anxious),
  "Mean"=wtd.mean(anxious,gross4),
  "Median"=wtd.quantile(anxious,gross4,probs=.5),
  "Std Deviation"=sqrt(wtd.var(anxious,gross4)),
  "Variance"=wtd.var(anxious,gross4),
  "Mode"=as.numeric(names(sort(-table(anxious)))[1]),
  "Range"=max(anxious)-min(anxious),
  "IQR"=wtd.quantile(anxious,gross4,probs=.75)-wtd.quantile(anxious,gross4,probs=.25),
  "Skewness"=Weighted.Desc.Stat::w.skewness(anxious,gross4),
```



```
"Kurtosis"=Weighted.Desc.Stat::w.kurtosis(anxious,gross4)
)),2)
```

```
##           [,1]      [,2]
## year      1920.00  2021.00
## Observations 12920.00 26259.00
## Mean         3.00    2.81
## Median       2.00    2.00
## Std Deviation 2.89    2.88
## Variance     8.34    8.31
## Mode         0.00    0.00
## Range        10.00   10.00
## IQR          5.00    5.00
## Skewness     0.65    0.76
## Kurtosis     -0.71   -0.57
```

Weighted quantiles

```
cbind(
  combine_years%>%filter(year==1920 & !is.na(anxious))%>%select(anxious,gross4)%>%summarise(
    "Quantiles"=
    round(wtd.quantile(anxious,gross4,probs=c(.01,.05,.1,.25,.5,.75,.9,.95,.99)),2)
  ),
  combine_years%>%filter(year==2021 & !is.na(anxious))%>%select(anxious,gross4)%>%summarise(
    "Quantiles"=
    round(wtd.quantile(anxious,gross4,probs=c(.01,.05,.1,.25,.5,.75,.9,.95,.99)),2)
  )
)
```

```
##   Quantiles Quantiles
## 1         0         0
## 2         0         0
## 3         0         0
## 4         0         0
## 5         2         2
## 6         5         5
## 7         7         7
## 8         8         8
## 9        10        10
```

Extreme values

```
cbind(
  combine_years%>%filter(year==1920 & !is.na(anxious))%>%select(anxious)%>%
  slice_min(order_by = anxious,n=5,with_ties = FALSE),
  combine_years%>%filter(year==2021 & !is.na(anxious))%>%select(anxious)%>%
  slice_max(order_by = anxious,n=5,with_ties = FALSE)
)
```

```
##   anxious anxious
## 1         0      10
## 2         0      10
## 3         0      10
## 4         0      10
## 5         0      10
```

Examines the distribution of life satisfaction in 2019-20 and 2020-21, including an assessment

of normality and discovery of outliers

```
round(t(combine_years%>%filter(!is.na(lifesat))%>%group_by(year)%>%summarise(
  "Observations"=length(lifesat),
  "Mean"=wtd.mean(lifesat,gross4),
  "Median"=wtd.quantile(lifesat,gross4,probs=.5),
  "Std Deviation"=sqrt(wtd.var(lifesat,gross4)),
  "Variance"=wtd.var(lifesat,gross4),
  "Mode"=as.numeric(names(sort(-table(lifesat)))[1]),
  "Range"=max(lifesat)-min(lifesat),
  "IQR"=wtd.quantile(lifesat,gross4,probs=.75)-wtd.quantile(lifesat,gross4,probs=.25),
  "Skewness"=Weighted.Desc.Stat::w.skewness(lifesat,gross4),
  "Kurtosis"=Weighted.Desc.Stat::w.kurtosis(lifesat,gross4)
)),2)
```

	[,1]	[,2]
## year	1920.00	2021.00
## Observations	12922.00	26266.00
## Mean	7.35	7.55
## Median	8.00	8.00
## Std Deviation	1.95	1.91
## Variance	3.79	3.65
## Mode	8.00	8.00
## Range	10.00	10.00
## IQR	3.00	2.00
## Skewness	-1.08	-1.27
## Kurtosis	1.65	2.28

Weighted quantiles

```
cbind(
  combine_years%>%filter(year==1920 & !is.na(lifesat))%>%select(lifesat,gross4)%>%summarise(
    "Quantiles"=
    round(wtd.quantile(lifesat,gross4,probs=c(.01,.05,.1,.25,.5,.75,.9,.95,.99)),2)
  ),
  combine_years%>%filter(year==1920 & !is.na(lifesat))%>%select(lifesat,gross4)%>%summarise(
    "Quantiles"=
    round(wtd.quantile(lifesat,gross4,probs=c(.01,.05,.1,.25,.5,.75,.9,.95,.99)),2)
  )
)
```

##	Quantiles	Quantiles
## 1	1	1
## 2	4	4
## 3	5	5
## 4	6	6
## 5	8	8
## 6	9	9
## 7	10	10
## 8	10	10
## 9	10	10

Extreme values

```
cbind(
  combine_years%>%filter(year==1920 & !is.na(lifesat))%>%select(lifesat)%>%
  slice_min(order_by = lifesat,n=5,with_ties = FALSE),
```

```
combine_years%>%filter(year==2021 & !is.na(lifesat))%>%select(lifesat)%>%
  slice_max(order_by = lifesat,n=5,with_ties = FALSE)
)
```

```
##   lifesat lifesat
## 1      0      10
## 2      0      10
## 3      0      10
## 4      0      10
## 5      0      10
```

Examines the distribution of meaning in 2019-20 and 2020-21, including an assessment of normality and discovery of outliers

```
round(t(combine_years%>%filter(!is.na(meaning))%>%group_by(year)%>%summarise(
  "Observations"=length(meaning),
  "Mean"=wtd.mean(meaning,gross4),
  "Median"=wtd.quantile(meaning,gross4,probs=.5),
  "Std Deviation"=sqrt(wtd.var(meaning,gross4)),
  "Variance"=wtd.var(meaning,gross4),
  "Mode"=as.numeric(names(sort(-table(meaning)))[1]),
  "Range"=max(meaning)-min(meaning),
  "IQR"=wtd.quantile(meaning,gross4,probs=.75)-wtd.quantile(meaning,gross4,probs=.25),
  "Skewness"=Weighted.Desc.Stat::w.skewness(meaning,gross4),
  "Kurtosis"=Weighted.Desc.Stat::w.kurtosis(meaning,gross4)
)),2)
```

```
##           [,1]      [,2]
## year      1920.00  2021.00
## Observations 12911.00 26219.00
## Mean         7.72    7.83
## Median        8.00    8.00
## Std Deviation  1.81    1.81
## Variance       3.28    3.28
## Mode          8.00    8.00
## Range        10.00   10.00
## IQR           2.00    2.00
## Skewness      -1.10   -1.23
## Kurtosis       1.83    2.33
```

Weighted quantiles

```
cbind(
  combine_years%>%filter(year==1920 & !is.na(meaning))%>%select(meaning,gross4)%>%summarise(
    "Quantiles"=
    round(wtd.quantile(meaning,gross4,probs=c(.01,.05,.1,.25,.5,.75,.9,.95,.99)),2)
  ),
  combine_years%>%filter(year==1920 & !is.na(meaning))%>%select(meaning,gross4)%>%summarise(
    "Quantiles"=
    round(wtd.quantile(meaning,gross4,probs=c(.01,.05,.1,.25,.5,.75,.9,.95,.99)),2)
  )
)
```

```
##   Quantiles Quantiles
## 1         2         2
## 2         5         5
```

```
## 3      5      5
## 4      7      7
## 5      8      8
## 6      9      9
## 7     10     10
## 8     10     10
## 9     10     10
```

Extreme values

```
cbind(
  combine_years%>%filter(year==1920 & !is.na(meaning))%>%select(meaning)%>%
  slice_min(order_by = meaning,n=5,with_ties = FALSE),
  combine_years%>%filter(year==2021 & !is.na(meaning))%>%select(meaning)%>%
  slice_max(order_by = meaning,n=5,with_ties = FALSE)
)
```

```
##      meaning meaning
## 1         0      10
## 2         0      10
## 3         0      10
## 4         0      10
## 5         0      10
```

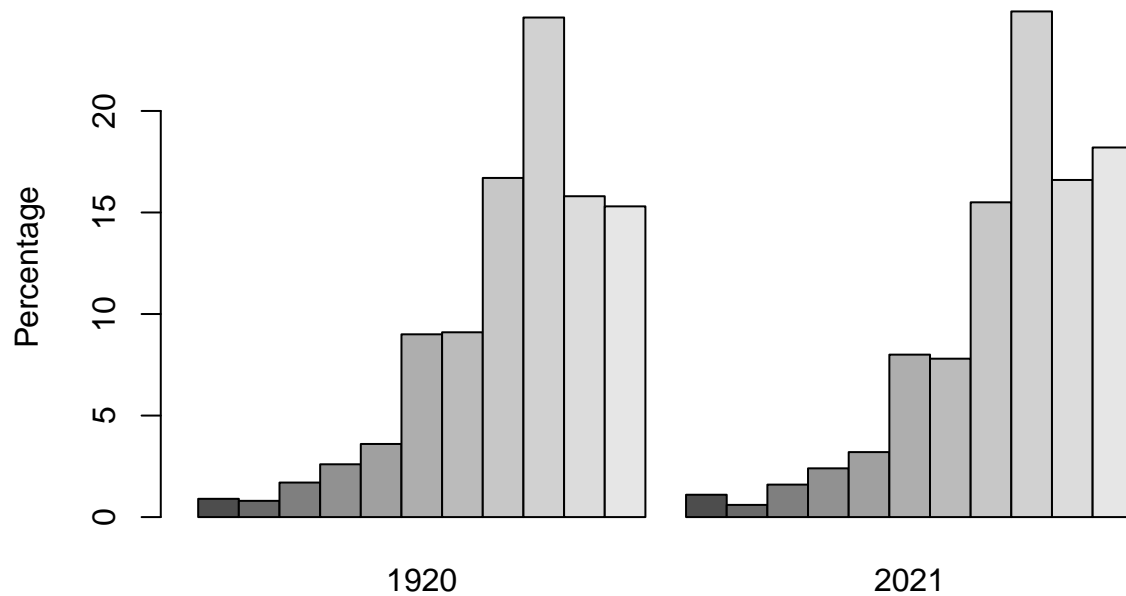
First getting the happiness tables (we could have reused
the frequency table we created above)

```
t<-
round(                                     ### Rounding result to 1 decimal
100*prop.table(                           ### Computing proportions, then multiplying by to get %
xtabs(~happywb+year, combine_years)      ### basic frequency table
,2)
,1)
```

The plotting

```
barplot(t, beside = T,
  main="Histogram to compare HAPPYWB in 2019-20 to 2020-21",
  xlab="How happy did you feel yesterday?",
  ylab="Percentage"
)
```

Histogram to compare HAPPYWB in 2019–20 to 2020–21



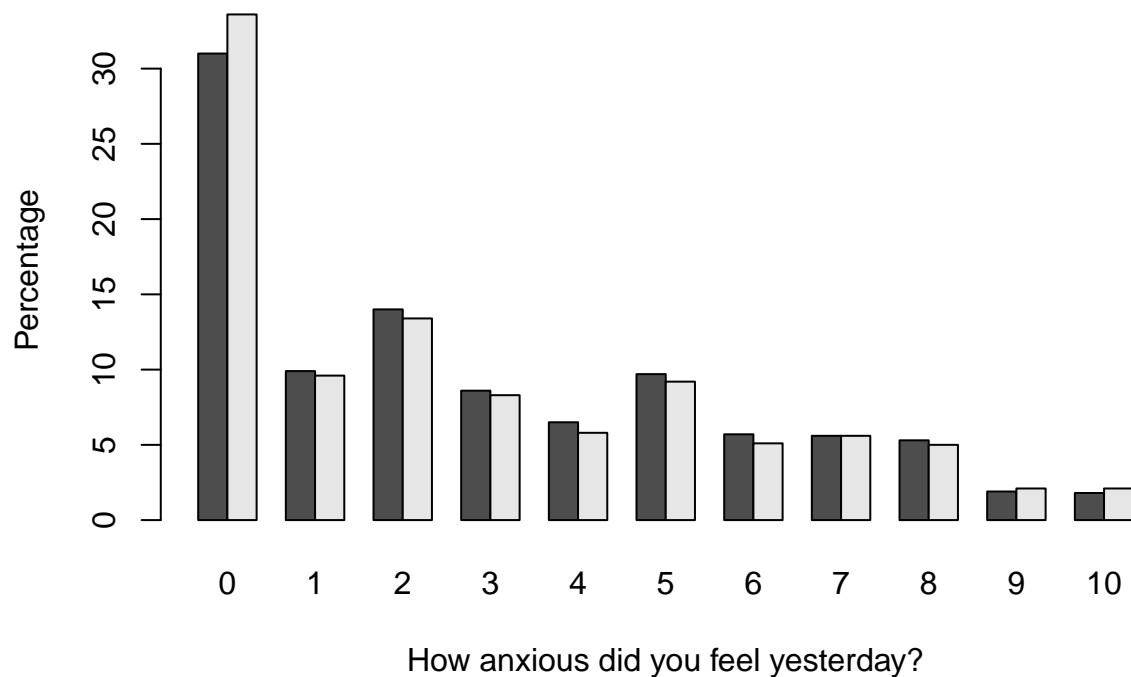
How happy did you feel yesterday?

```
### First getting the anxiety table
t<-
round(
  100*prop.table(
    xtabs(~year+anxious, combine_years)
  ,1)
,1)

### Rounding result to 1 decimal
### Computing proportions, then multiplying by to get %
### basic frequency table

### The actual plotting
barplot(t, beside = T,
  main="Histogram to compare ANXIOUS in 2019-20 to 2020-21",
  xlab="How anxious did you feel yesterday?",
  ylab="Percentage"
)
```

Histogram to compare ANXIOUS in 2019–20 to 2020–21

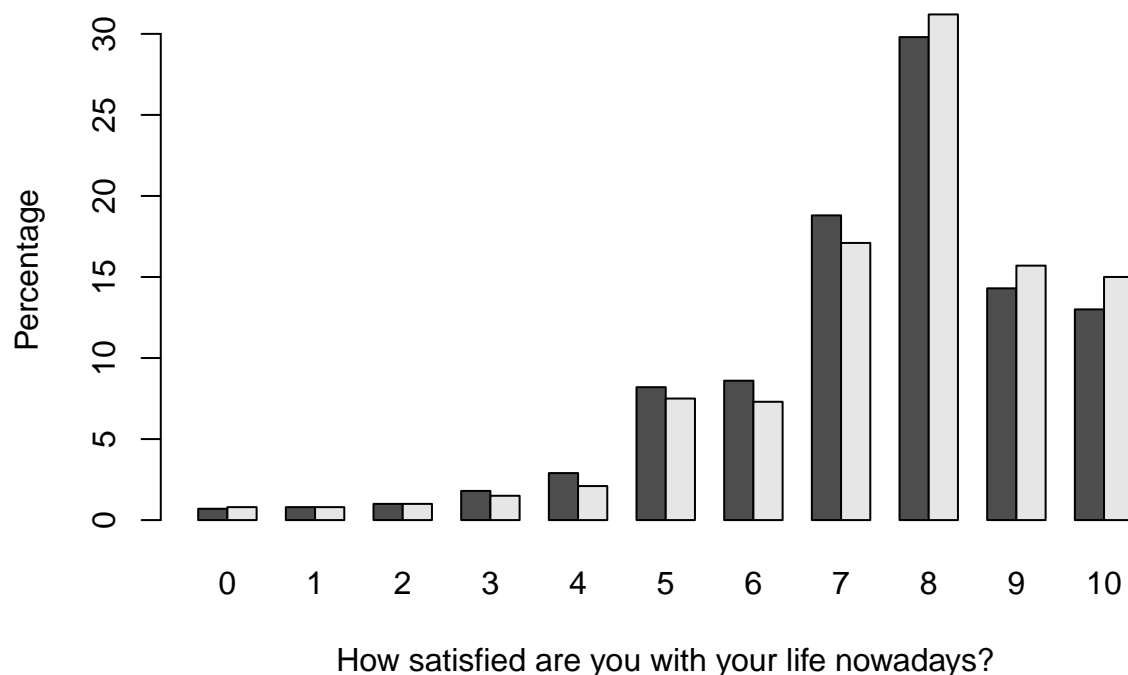


```
### Getting the life satisfaction table
t<-
round(
  100*prop.table(
    xtabs(~year+lifesat, combine_years)
  ,1)
,1)

### Rounding result to 1 decimal
### Computing proportions, then multiplying by to get %
### basic frequency table

### The actual plotting
barplot(t, beside = T,
  main="Histogram to compare LIFESAT in 2019-20 to 2020-21",
  xlab="How satisfied are you with your life nowadays?",
  ylab="Percentage"
)
```

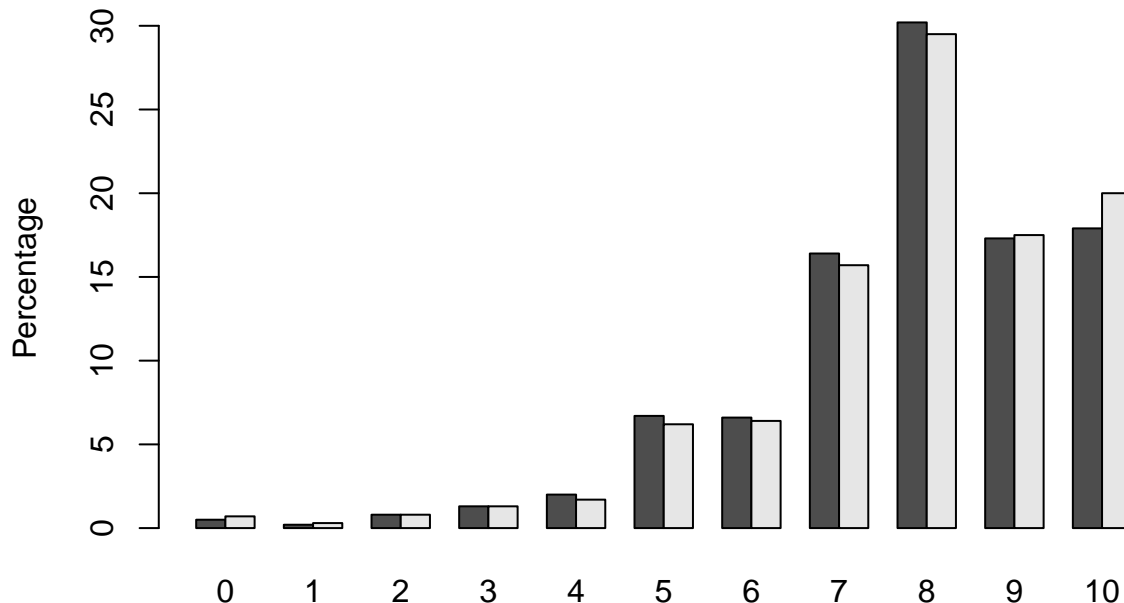
Histogram to compare LIFESAT in 2019–20 to 2020–21



```
### Getting the life meaningfulness data
t<-
round(
  100*prop.table(
    xtabs(~year+meaning, combine_years)
  ,1)
,1)
### Rounding result to 1 decimal
### Computing proportions, then multiplying by to get %
### basic frequency table

### The actual plotting
barplot(t, beside = T,
  main="Histogram to compare MEANING in 2019-20 to 2020-21",
  xlab="To what extent do you feel that things you do in your life have meaning?",
  ylab="Percentage"
)
```

Histogram to compare MEANING in 2019–20 to 2020–21



To what extent do you feel that things you do in your life have meaning?

Provides grossed analysis of variance and Wilcoxon Scores (Rank Sums) for wellbeing variables by year

4.1.1 ANOVA happiness by year

ANOVA amounts to running a statistical test of significance of differences in the mean values of a variable across two or more groups. In most cases, linear regression with a single categorical independent (ie X) variable is identical. We will be choosing this approach here, as it is easier to implement in R.

```
summary(
  lm(happywb~as.factor(year),data=combine_years,weights=gross4))

##
## Call:
## lm(formula = happywb ~ as.factor(year), data = combine_years,
##     weights = gross4)
##
## Weighted Residuals:
##      Min       1Q   Median       3Q      Max
## -743.70  -36.56   19.29   60.68  463.14
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      7.29116    0.01536  474.685  <2e-16 ***
## as.factor(year)2021  0.19046    0.02146   8.875  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 93.6 on 39183 degrees of freedom
## (11638 observations deleted due to missingness)
## Multiple R-squared:  0.002006,    Adjusted R-squared:  0.001981
```



```
## F-statistic: 78.77 on 1 and 39183 DF, p-value: < 2.2e-16
### We need to use the Survey package in order to access the weighted rank test functions
### (Wilcoxon/Kruskal-Wallis)
library(survey)

### We also need to declare the survey design, in this case we will only specify the
### grossing weights
design <- svydesign(ids = ~0, data=combine_years,
weights=combine_years$gross4)

svyranktest(happywb ~ as.factor(year), design = design)

##
## Design-based KruskalWallis test
##
## data: happywb ~ as.factor(year)
## t = 7.5714, df = 39183, p-value = 3.773e-14
## alternative hypothesis: true difference in mean rank score is not equal to 0
## sample estimates:
## difference in mean rank score
## 0.02985664
```

4.1.2 ANOVA Anxiety amount by year

/Provides grossed analysis of variance and Kruskal-Wallis Rank Sums for State Pension benefit amounts by year/

```
summary(lm(anxious~as.factor(year),data=combine_years,weights=gross4))

##
## Call:
## lm(formula = anxious ~ as.factor(year), data = combine_years,
## weights = gross4)
##
## Weighted Residuals:
## Min 1Q Median 3Q Max
## -517.17 -99.59 -31.00 79.40 747.41
##
## Coefficients:
## Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2.99605 0.02088 143.491 < 2e-16 ***
## as.factor(year)2021 -0.18385 0.02917 -6.303 2.95e-10 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 127.2 on 39177 degrees of freedom
## (11644 observations deleted due to missingness)
## Multiple R-squared: 0.001013, Adjusted R-squared: 0.0009875
## F-statistic: 39.73 on 1 and 39177 DF, p-value: 2.951e-10

svyranktest(anxious ~ as.factor(year), design = design)

##
## Design-based KruskalWallis test
##
## data: anxious ~ as.factor(year)
```

```
## t = -5.1759, df = 39177, p-value = 2.279e-07
## alternative hypothesis: true difference in mean rank score is not equal to 0
## sample estimates:
## difference in mean rank score
## -0.02072376
```

4.1.3 ANOVA Life satisfaction by year

/Provides grossed analysis of variance and Kruskal-Wallis Rank Sums for Personal Independence Payment (PIP) Daily Living benefit amounts by year/

```
summary(lm(anxious~as.factor(year),data=combine_years,weights=gross4))
```

```
##
## Call:
## lm(formula = anxious ~ as.factor(year), data = combine_years,
##     weights = gross4)
##
## Weighted Residuals:
##      Min       1Q   Median       3Q      Max
## -517.17  -99.59  -31.00   79.40  747.41
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      2.99605     0.02088 143.491 < 2e-16 ***
## as.factor(year)2021 -0.18385     0.02917  -6.303 2.95e-10 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 127.2 on 39177 degrees of freedom
## (11644 observations deleted due to missingness)
## Multiple R-squared:  0.001013,    Adjusted R-squared:  0.0009875
## F-statistic: 39.73 on 1 and 39177 DF,  p-value: 2.951e-10
```

```
svyranktest(anxious ~ as.factor(year), design = design)
```

```
##
## Design-based KruskalWallis test
##
## data:  anxious ~ as.factor(year)
## t = -5.1759, df = 39177, p-value = 2.279e-07
## alternative hypothesis: true difference in mean rank score is not equal to 0
## sample estimates:
## difference in mean rank score
## -0.02072376
```

4.1.4 ANOVA life meaning by year

/Provides grossed analysis of variance and Kruskal-Wallis Rank Sums for Personal Independence Payment (PIP) Mobility benefit amounts by year/

```
summary(lm(anxious~as.factor(year),data=combine_years,weights=gross4))
```

```
##
## Call:
## lm(formula = anxious ~ as.factor(year), data = combine_years,
##     weights = gross4)
```

```
##
## Weighted Residuals:
##      Min      1Q  Median      3Q      Max
## -517.17  -99.59  -31.00   79.40  747.41
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      2.99605    0.02088 143.491 < 2e-16 ***
## as.factor(year)2021 -0.18385    0.02917  -6.303 2.95e-10 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 127.2 on 39177 degrees of freedom
## (11644 observations deleted due to missingness)
## Multiple R-squared:  0.001013, Adjusted R-squared:  0.0009875
## F-statistic: 39.73 on 1 and 39177 DF, p-value: 2.951e-10
svyranktest(anxious ~ as.factor(year), design = design)
```

```
##
## Design-based KruskalWallis test
##
## data:  anxious ~ as.factor(year)
## t = -5.1759, df = 39177, p-value = 2.279e-07
## alternative hypothesis: true difference in mean rank score is not equal to 0
## sample estimates:
## difference in mean rank score
##                -0.02072376
```

4.2 Tables of wellbeing variable scores against other variables

Create a grossed table of life satisfaction score against whether the adult has difficulty with mental health in 2019-20

```
### 2019-20
round(
  100*prop.table(
    xtabs(gross4~lifesat+disd07,combine_years%>%filter(year==1920))
    ,2),1)
```

```
##      disd07
## lifesat    1    2    3
##    0    3.4  1.0  0.6
##    1    1.7  0.9  0.9
##    2    4.8  1.0  0.3
##    3    7.3  1.9  0.7
##    4    8.3  3.6  3.1
##    5   18.7 11.3  3.6
##    6   13.4  9.8  8.1
##    7   18.8 18.2 19.3
##    8   14.1 28.3 32.0
##    9    5.8 11.8 15.9
##   10    3.8 12.1 15.4
```

```
### 2020-21
round(
```

```
100*prop.table(
  xtabs(gross4~lifesat+disd07,combine_years%>%filter(year==2021))
,2),1)
```

```
##          disd07
## lifesat    1    2    3
##    0    4.1  1.0  0.4
##    1    2.7  0.8  0.2
##    2    4.4  1.4  0.4
##    3    6.9  1.7  0.6
##    4    7.8  2.4  1.3
##    5   16.7 10.6  4.6
##    6   13.4  8.2  6.0
##    7   17.6 16.7 15.3
##    8   16.2 29.3 35.3
##    9    5.9 13.9 17.7
##   10    4.3 14.1 18.1
```

Create a grossed table of anxiousness score against the household income band of the adult in 2019-20

```
### 2019-20
round(
100*prop.table(
  xtabs(gross4~anxious+hhincbnd,combine_years%>%filter(year==1920))
,2),1)
```

```
##          hhincbnd
## anxious    1    2    3    4    5    6    7    8    9   10   11
##    0   27.3 32.9 29.6 33.7 31.3 31.3 31.2 25.9 22.9 27.6 27.5
##    1    6.9  6.9  8.3  8.9 10.9  8.7 10.5  8.1 13.8 12.5 10.9
##    2   12.7 10.6 13.2 14.4 14.3 13.3 17.3 16.5 16.5 13.1 16.4
##    3   10.6  7.1  9.6  5.5  8.9  9.8  8.2 10.9  7.7 13.7  9.4
##    4    6.6  6.3  5.6  7.0  6.0  6.9  6.6  8.5  6.2  5.2  5.1
##    5   13.7 11.8 11.0  9.3  9.2  9.5  8.9  9.1  9.4 10.7  9.6
##    6    4.0  5.3  5.7  6.6  5.9  5.3  5.3  6.4  5.7  4.0  9.0
##    7    6.1  6.3  5.4  5.2  5.1  7.0  6.4  5.9  7.5  6.8  5.0
##    8    7.1  7.0  6.1  5.4  4.4  5.9  3.2  5.5  5.6  2.4  4.1
##    9    2.9  3.3  2.9  1.7  1.9  0.8  2.2  2.2  1.7  3.0  1.3
##   10    2.1  2.5  2.6  2.2  2.0  1.5  0.1  1.0  2.9  1.0  1.6
```

```
### 2020-21
round(
100*prop.table(
  xtabs(gross4~anxious+hhincbnd,combine_years%>%filter(year==2021))
,2),1)
```

```
##          hhincbnd
## anxious    1    2    3    4    5    6    7    8    9   10   11
##    0   31.4 33.7 34.2 34.2 33.8 34.7 33.4 35.5 32.0 31.4 33.6
##    1    7.7  6.7  8.7  8.8  9.8 10.3 11.6 12.1 11.5 10.8 13.0
##    2   12.9 12.5 12.2 12.6 14.0 13.9 13.1 16.0 12.1 12.9 14.3
##    3    6.9  7.9  8.0  8.6  8.8  8.2  8.8  7.6 10.5 11.0  9.0
##    4    6.8  5.5  5.7  5.8  5.9  5.9  6.3  4.6  7.1  7.1  6.4
##    5    8.4 10.6  9.8 10.3  9.1  9.4  9.0  8.9  9.5  8.0  7.3
##    6    5.9  4.9  4.9  5.2  5.5  4.5  4.9  4.5  6.3  5.5  5.5
##    7    6.2  6.0  5.6  5.7  5.6  5.2  4.9  5.5  5.6  6.2  4.9
```

##	8	7.2	6.4	5.8	5.1	4.5	5.4	4.3	3.0	2.1	3.7	4.0
##	9	2.5	2.8	2.4	1.9	1.5	1.5	1.9	1.7	2.4	2.1	1.5
##	10	4.1	3.1	2.8	1.9	1.4	1.1	1.7	0.7	0.9	1.3	0.6

5 Rapid analysis

5.1 FRS households by tenure between 2018-19 and 2020-21

```
setwd("/home/piet/Dropbox/work/UKDS/FRS workshop/" )
tenure1<-read.spss("UKDA-8948-spss/spss/spss25/househol.sav",to.data.frame = T,
use.value.labels = T,
add.undeclared.levels="append")%>%select(SERNUM,TENURE,GROSS4)
names(tenure1)<-tolower(names(tenure1))

tenure2<-read.spss("UKDA-8802-spss/spss/spss25/househol.sav",to.data.frame = T,
use.value.labels = T,
add.undeclared.levels="append")%>%select(SERNUM,TENURE,GROSS4)
names(tenure2)<-tolower(names(tenure2))

tenure3<-read.spss("UKDA-8633-spss/househol.sav",to.data.frame = T,
use.value.labels = T,
add.undeclared.levels="append")%>%select(sernum,TENURE,GROSS4)
names(tenure3)<-tolower(names(tenure3))
```

2020-21

```
tmptab<-xtabs(gross4~tenure,tenure1,drop.unused.levels=T)
cbind(
"Freq"= tmptab,
"Pct"=round(100*prop.table(tmptab),1),
"Cum. freq."=cumsum(tmptab),
"Cum. pct"= round(cumsum(100*prop.table(tmptab)),2)
)
```

##	Freq	Pct	Cum. freq.	Cum. pct
## Owns it outright	10228059	36.3	10228059	36.26
## Buying with the help of a mortgage	7879072	27.9	18107131	64.20
## Part own, part rent	125263	0.4	18232394	64.64
## Rents	9715232	34.4	27947626	99.08
## Rent-free	258704	0.9	28206330	100.00

2019-20

```
tmptab<-xtabs(gross4~tenure,tenure2,drop.unused.levels=T)
cbind(
"Freq"= tmptab,
"Pct"=round(100*prop.table(tmptab),1),
"Cum. freq."=cumsum(tmptab),
"Cum. pct"= round(cumsum(100*prop.table(tmptab)),2)
)
```

##	Freq	Pct	Cum. freq.	Cum. pct
## Owns it outright	9952274	35.5	9952274	35.51
## Buying with the help of a mortgage	7849704	28.0	17801978	63.52
## Part own, part rent	187136	0.7	17989114	64.19
## Rents	9759793	34.8	27748907	99.02
## Rent-free	270662	1.0	28019569	99.98
## Squatting	4242	0.0	28023811	100.00

2018-19

```

tmptab<-xtabs(gross4~tenure,tenure3,drop.unused.levels=T)
cbind(
  "Freq"= tmptab,
  "Pct"=round(100*prop.table(tmptab),1),
  "Cum. freq."=cumsum(tmptab),
  "Cum. pct"= round(cumsum(100*prop.table(tmptab)),2)
)

```

##	Freq	Pct	Cum. freq.	Cum. pct
## Owns it outright	9792127	35.2	9792127	35.19
## Buying with the help of a mortgage	7758708	27.9	17550835	63.07
## Part own, part rent	139479	0.5	17690314	63.57
## Rents	9889994	35.5	27580308	99.11
## Rent-free	245099	0.9	27825407	99.99
## Squatting	3380	0.0	27828787	100.00

5.2 FRS households by Council Tax Band between 2018-19 and 2020-21

```

setwd("/home/piet/Dropbox/work/UKDS/FRS workshop/" )
ctband1<-read.spss("UKDA-8948-spss/spss/spss25/househol.sav",to.data.frame = T,
  use.value.labels = T,
  add.undeclared.levels="append")%>%select(SERNUM,CTBAND,GROSS4)
names(ctband1)<-tolower(names(ctband1))

ctband2<-read.spss("UKDA-8802-spss/spss/spss25/househol.sav",to.data.frame = T,
  use.value.labels = T,
  add.undeclared.levels="append")%>%select(SERNUM,CTBAND,GROSS4)
names(ctband2)<-tolower(names(ctband2))

ctband3<-read.spss("UKDA-8633-spss/househol.sav",to.data.frame = T,
  use.value.labels = T,
  add.undeclared.levels="append")%>%select(sernum,CTBAND,GROSS4)
names(ctband3)<-tolower(names(ctband3))

```

Gives the grossed frequency, percentage, cumulative frequency and cumulative percentage of each category for the variable 'ctband' in 2020-21

```

tmptab<-xtabs(gross4~ctband,ctband1,drop.unused.levels=T)
cbind(
  "Freq"= tmptab,
  "Pct"=round(100*prop.table(tmptab),1),
  "Cum. freq."=cumsum(tmptab),
  "Cum. pct"= round(cumsum(100*prop.table(tmptab)),2)
)

```

##	Freq	Pct	Cum. freq.	Cum. pct
## Band A	6310843	23.0	6310843	22.98
## Band B	5461981	19.9	11772824	42.87
## Band C	5583009	20.3	17355833	63.19
## Band D	4469946	16.3	21825779	79.47
## Band E	2932644	10.7	24758423	90.15
## Band F	1557968	5.7	26316391	95.82
## Band G	870044	3.2	27186435	98.99
## Band H	122557	0.4	27308992	99.43
## Band I	1262	0.0	27310254	99.44

```
## Household not valued separately    154516    0.6    27464770    100.00
```

Gives the grossed frequency, percentage, cumulative frequency and cumulative percentage of each category for the variable 'ctband' in 2019-20

```
tmptab<-xtabs(gross4~ctband,ctband2,drop.unused.levels=T)
cbind(
  "Freq"= tmptab,
  "Pct"=round(100*prop.table(tmptab),1),
  "Cum. freq."=cumsum(tmptab),
  "Cum. pct"= round(cumsum(100*prop.table(tmptab)),2)
)
```

##		Freq	Pct	Cum. freq.	Cum. pct
## Band A		6198194	22.7	6198194	22.72
## Band B		5419568	19.9	11617762	42.58
## Band C		5525309	20.2	17143071	62.83
## Band D		4426141	16.2	21569212	79.05
## Band E		2756851	10.1	24326063	89.15
## Band F		1512329	5.5	25838392	94.69
## Band G		1028904	3.8	26867296	98.46
## Band H		112002	0.4	26979298	98.88
## Band I		5872	0.0	26985170	98.90
## Household not valued separately		300984	1.1	27286154	100.00

Gives the grossed frequency, percentage, cumulative frequency and cumulative percentage of each category for the variable 'ctband' in 2018-19

```
tmptab<-xtabs(gross4~ctband,ctband3,drop.unused.levels=T)
cbind(
  "Freq"= tmptab,
  "Pct"=round(100*prop.table(tmptab),1),
  "Cum. freq."=cumsum(tmptab),
  "Cum. pct"= round(cumsum(100*prop.table(tmptab)),2)
)
```

##		Freq	Pct	Cum. freq.	Cum. pct
## Band A		6188935	22.8	6188935	22.84
## Band B		5394817	19.9	11583752	42.75
## Band C		5418607	20.0	17002359	62.75
## Band D		4443995	16.4	21446354	79.15
## Band E		2669700	9.9	24116054	89.00
## Band F		1558186	5.8	25674240	94.76
## Band G		1002356	3.7	26676596	98.46
## Band H		121146	0.4	26797742	98.90
## Band I		5918	0.0	26803660	98.92
## Household not valued separately		291550	1.1	27095210	100.00

5.3 FRS adult respondents by highest level of qualification between 2018-19 and 2020-21

```
setwd("/home/piet/Dropbox/work/UKDS/FRS workshop/" )
dvhiqual12<-rbind(
  cbind(read.spss("UKDA-8948-spss/spss/spss25/adult.sav",to.data.frame = T,
    use.value.labels = F,
    add.undeclared.levels="append"))%>%select(SERNUM,BENUNIT,PERSON, DVHIQUAL,GROSS4),
```



```

year=2021),
cbind(read.spss("UKDA-8802-spss/spss25/adult.sav",to.data.frame = T,
use.value.labels = F,
add.undeclared.levels="append")%>%select(SERNUM,BENUNIT,PERSON, DVHIQUAL,GROSS4),
year=2020)
)
names(dvhiqual12)<-tolower(names(dvhiqual12))

### Unfortunately one of the datasets has a different variable name, so we need to deal
### with it separately
dvhiqual3<-cbind(read.spss("UKDA-8633-spss/adult.sav",to.data.frame = T,
use.value.labels = F,
add.undeclared.levels="append")%>%select(sernum,BENUNIT,PERSON, DVHIQUAL,GROSS4),
year=2019)
names(dvhiqual3)<-tolower(names(dvhiqual3))

dvhiqual123<-rbind(
  dvhiqual12,dvhiqual3
)

dvhiqual123<-dvhiqual123%>%mutate(dvhiqual.s=case_when(
dvhiqual>=1 & dvhiqual<=16 ~ 'Level 4 or greater',
dvhiqual>=17 & dvhiqual<=34 ~ 'Level 3',
dvhiqual>=35 & dvhiqual<=54 ~ 'Level 2',
dvhiqual>=55 & dvhiqual<=82 ~ 'Level 1',
dvhiqual>=83 & dvhiqual<=85 ~ 'Entry level',
dvhiqual==86 ~ 'Other',
is.na(dvhiqual==T) ~ 'No reported qualification'
))

### Percentages
cbind(
round(100*prop.table(xtabs(gross4~dvhiqual.s+year,dvhiqual123),2),1),
### Then grossed frequencies
xtabs(gross4~dvhiqual.s+year,dvhiqual123)
)

```

##	2019	2020	2021	2019	2020	2021
## Entry level	0.6	0.7	0.4	306778	338955	183998
## Level 1	8.7	8.6	10.2	4504197	4424985	5289459
## Level 2	12.0	11.6	13.0	6191677	6012687	6742296
## Level 3	18.1	18.7	19.5	9347694	9644517	10059787
## Level 4 or greater	42.6	43.3	40.8	21972516	22390103	21111412
## No reported qualification	15.3	14.5	13.2	7871914	7477719	6844681
## Other	2.6	2.7	2.9	1359183	1370074	1480014