

Modeller

MSB 105 - Assignment 4

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```
suppressPackageStartupMessages({  
  library(tidyverse)  
  library(lubridate)  
  library(modelr)  
  library(broom)  
  library(lmtest)  
  library(sandwich)  
  library(viridis)  
})
```

Henter csv. filen:

```
pm2 <- read_csv("data/pm2.csv", show_col_types = FALSE)
```

```
## New names:  
## * ' -> ...1
```

Muterer:

```
pm2 <- pm2 %>%  
  mutate(  
    fnr = str_sub(knr, 1,2),  
    aar_f = str_sub(aar)  
  )
```

```
head(pm2)
```

```
## # A tibble: 6 x 19  
##   ...1 knr    aar knavn    pm2 Menn_ya_p Kvinner_ya_p Total_ya_p inc_k1 inc_k5  
##   <dbl> <chr> <dbl> <chr>  <dbl>    <dbl>        <dbl>    <dbl> <dbl> <dbl>  
## 1     1 0101   2008 Halden 13427     59.7         56.8      58.3  24.5  13.6  
## 2     2 0101   2009 Halden 13095     59.8         57.0      58.4  24.4  14.1  
## 3     3 0101   2010 Halden 13832     59.6         57.1      58.3  23.9  13.7  
## 4     4 0101   2011 Halden 14915     59.8         57.2      58.5  24    14  
## 5     5 0101   2012 Halden 15473     59.5         57.0      58.2  23.9  14  
## 6     6 0101   2013 Halden 15461     59.0         56.7      57.9  24.1  13.4  
## # ... with 9 more variables: uni_k_mf <dbl>, uni_k_m <dbl>, uni_k_f <dbl>,  
## #   uni_l_mf <dbl>, uni_l_m <dbl>, uni_l_f <dbl>, Trade_p <dbl>, fnr <chr>,  
## #   aar_f <chr>
```

parse_factor funksjonen:

```
pm2 %>%
  mutate(
    fnr = parse_factor(fnr, levels = fnr),
    aar_f = parse_factor(aar_f, levels = aar_f)
  )
```

```
## # A tibble: 2,140 x 19
##   ...1 knr      aar knavn      pm2 Menn_ya_p Kvinner_ya_p Total_ya_p inc_k1 inc_k5
##   <dbl> <chr> <dbl> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1     1 0101  2008 Halden 13427    59.7    56.8    58.3  24.5  13.6
## 2     2 0101  2009 Halden 13095    59.8    57.0    58.4  24.4  14.1
## 3     3 0101  2010 Halden 13832    59.6    57.1    58.3  23.9  13.7
## 4     4 0101  2011 Halden 14915    59.8    57.2    58.5  24    14
## 5     5 0101  2012 Halden 15473    59.5    57.0    58.2  23.9  14
## 6     6 0101  2013 Halden 15461    59.0    56.7    57.9  24.1  13.4
## 7     7 0101  2014 Halden 17164    58.8    56.7    57.7  23.9  13.5
## 8     8 0101  2015 Halden 17427    58.7    56.8    57.8  24    13.7
## 9     9 0101  2016 Halden 18941    58.7    56.6    57.7  24    13.8
## 10    10 0101  2017 Halden 20143    58.9    56.9    57.9  23.7  14
## # ... with 2,130 more rows, and 9 more variables: uni_k_mf <dbl>,
## #   uni_k_m <dbl>, uni_k_f <dbl>, uni_l_mf <dbl>, uni_l_m <dbl>, uni_l_f <dbl>,
## #   Trade_p <dbl>, fnr <fct>, aar_f <fct>
```

muterer:

```
pm2 <- pm2 %>%
  mutate(
    Trade_pc_100K = Trade_p/100000
  )
```

```
head(pm2, n = 4)
```

```
## # A tibble: 4 x 20
##   ...1 knr      aar knavn      pm2 Menn_ya_p Kvinner_ya_p Total_ya_p inc_k1 inc_k5
##   <dbl> <chr> <dbl> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1     1 0101  2008 Halden 13427    59.7    56.8    58.3  24.5  13.6
## 2     2 0101  2009 Halden 13095    59.8    57.0    58.4  24.4  14.1
## 3     3 0101  2010 Halden 13832    59.6    57.1    58.3  23.9  13.7
## 4     4 0101  2011 Halden 14915    59.8    57.2    58.5  24    14
## # ... with 10 more variables: uni_k_mf <dbl>, uni_k_m <dbl>, uni_k_f <dbl>,
## #   uni_l_mf <dbl>, uni_l_m <dbl>, uni_l_f <dbl>, Trade_p <dbl>, fnr <chr>,
## #   aar_f <chr>, Trade_pc_100K <dbl>
```

Modell

```
mod1 <- 'pm2 ~ aar_f + Total_ya_p + inc_k1 + inc_k5 + uni_k_mf + uni_l_mf + Trade_pc_100K'
```

i.

```
lm1 <- lm(mod1, data = pm2, subset = complete.cases(pm2))
```

```
summary(lm1)
```

```
##
## Call:
## lm(formula = mod1, data = pm2, subset = complete.cases(pm2))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -8516.6 -1472.1   -29.9   1467.3 15736.3
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -20400.74    2663.02  -7.661 2.79e-14 ***
## aar_f2009      104.15     244.77   0.426 0.670512
## aar_f2010      908.13     245.16   3.704 0.000217 ***
## aar_f2011     1663.93     245.86   6.768 1.68e-11 ***
## aar_f2012     2240.48     247.10   9.067 < 2e-16 ***
## aar_f2013     2869.30     248.31  11.555 < 2e-16 ***
## aar_f2014     2863.22     250.54  11.428 < 2e-16 ***
## aar_f2015     3525.22     253.08  13.929 < 2e-16 ***
## aar_f2016     4274.99     255.81  16.711 < 2e-16 ***
## aar_f2017     5146.33     258.50  19.909 < 2e-16 ***
## Total_ya_p      582.44      38.94  14.957 < 2e-16 ***
## inc_k1        -376.99      30.29 -12.445 < 2e-16 ***
## inc_k5         194.35      22.87   8.498 < 2e-16 ***
## uni_k_mf       -82.02      29.42  -2.788 0.005357 **
## uni_l_mf      1206.86      42.22  28.585 < 2e-16 ***
## Trade_pc_100K   871.99     218.42   3.992 6.77e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2531 on 2124 degrees of freedom
## Multiple R-squared:  0.8346, Adjusted R-squared:  0.8334
## F-statistic: 714.3 on 15 and 2124 DF,  p-value: < 2.2e-16
```

ii. Legger til residualer:

```
pm2 %>%
```

```
  add_residuals(lm1)
```

```
## # A tibble: 2,140 x 21
##   ...1 knr    aar knavn    pm2 Menn_ya_p Kvinner_ya_p Total_ya_p inc_k1 inc_k5
##   <dbl> <chr> <dbl> <chr> <dbl>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>
## 1     1 0101   2008 Halden 13427    59.7      56.8      58.3     24.5     13.6
## 2     2 0101   2009 Halden 13095    59.8      57.0      58.4     24.4     14.1
## 3     3 0101   2010 Halden 13832    59.6      57.1      58.3     23.9     13.7
## 4     4 0101   2011 Halden 14915    59.8      57.2      58.5     24      14
## 5     5 0101   2012 Halden 15473    59.5      57.0      58.2     23.9     14
## 6     6 0101   2013 Halden 15461    59.0      56.7      57.9     24.1     13.4
## 7     7 0101   2014 Halden 17164    58.8      56.7      57.7     23.9     13.5
```

```
## 8      8 0101    2015 Halden 17427      58.7      56.8      57.8    24      13.7
## 9      9 0101    2016 Halden 18941      58.7      56.6      57.7    24      13.8
## 10     10 0101    2017 Halden 20143      58.9      56.9      57.9   23.7    14
## # ... with 2,130 more rows, and 11 more variables: uni_k_mf <dbl>,
## #   uni_k_m <dbl>, uni_k_f <dbl>, uni_l_mf <dbl>, uni_l_m <dbl>, uni_l_f <dbl>,
## #   Trade_p <dbl>, fnr <chr>, aar_f <chr>, Trade_pc_100K <dbl>, resid <dbl>
```

```
head(pm2, n = 4)
```

```
## # A tibble: 4 x 20
##   ...1 knr      aar knavn    pm2 Menn_ya_p Kvinner_ya_p Total_ya_p inc_k1 inc_k5
##   <dbl> <chr> <dbl> <chr> <dbl>      <dbl>      <dbl>      <dbl> <dbl> <dbl>
## 1     1 0101    2008 Halden 13427      59.7      56.8      58.3    24.5    13.6
## 2     2 0101    2009 Halden 13095      59.8      57.0      58.4    24.4    14.1
## 3     3 0101    2010 Halden 13832      59.6      57.1      58.3    23.9    13.7
## 4     4 0101    2011 Halden 14915      59.8      57.2      58.5     24     14
## # ... with 10 more variables: uni_k_mf <dbl>, uni_k_m <dbl>, uni_k_f <dbl>,
## #   uni_l_mf <dbl>, uni_l_m <dbl>, uni_l_f <dbl>, Trade_p <dbl>, fnr <chr>,
## #   aar_f <chr>, Trade_pc_100K <dbl>
```

i.

Man leser ut gjennomsnittlig kvadratmeterpris for en enebolig ($pm2$) for de forskjellige årene. Vi ser at $pm2$ stiger jevnt og trutt.

ii.

Vi vil si at fortegnene er som forventet. Dersom vi har tolket modellen riktig, så vil $pm2$ være mindre for dem nederste kvintilen (inc_k1) enn for den øverste (inc_k5). Det samme gjelder for de med kort og lang utdanning.

Dette er nok fordi den rikere delen av befolkninge, og de med høyere utdanning, sannsynligvis har mer attraktive eneboliger som gjør at $pm2$ er høyere.

Heteroskedastisitet

i.

```
bptest(lm1)
```

```
##
## studentized Breusch-Pagan test
##
## data:  lm1
## BP = 352.89, df = 15, p-value < 2.2e-16
```

ii.

Veldig høy p-verdi. Da kan H_0 forkastes og vi kan med sterke bevis si at det foreligger Heteroskedastisitet.

iii.

```
coeftest(lm1)
```

```
##
## t test of coefficients:
##
##          Estimate Std. Error  t value  Pr(>|t|)
## (Intercept) -20400.742   2663.022  -7.6607 2.790e-14 ***
## aar_f2009     104.150    244.767   0.4255 0.6705118
## aar_f2010     908.129    245.156   3.7043 0.0002174 ***
## aar_f2011    1663.926    245.857   6.7679 1.685e-11 ***
## aar_f2012    2240.475    247.095   9.0672 < 2.2e-16 ***
## aar_f2013    2869.297    248.315  11.5551 < 2.2e-16 ***
## aar_f2014    2863.224    250.537  11.4283 < 2.2e-16 ***
## aar_f2015    3525.223    253.083  13.9291 < 2.2e-16 ***
## aar_f2016    4274.990    255.812  16.7114 < 2.2e-16 ***
## aar_f2017    5146.326    258.498  19.9086 < 2.2e-16 ***
## Total_ya_p     582.436     38.941  14.9568 < 2.2e-16 ***
## inc_k1       -376.989     30.291 -12.4455 < 2.2e-16 ***
## inc_k5        194.354     22.871   8.4979 < 2.2e-16 ***
## uni_k_mf      -82.023     29.424  -2.7876 0.0053574 **
## uni_l_mf     1206.857     42.219  28.5853 < 2.2e-16 ***
## Trade_pc_100K  871.993    218.422   3.9922 6.768e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
vcovHC(lm1)
```

```
##          (Intercept)    aar_f2009    aar_f2010    aar_f2011    aar_f2012
## (Intercept)  9297989.37 -26519.17426 -34751.3931 -64358.9799 -88195.7750
## aar_f2009    -26519.17  42579.51052  22306.6988  22379.0191  22461.1963
## aar_f2010    -34751.39  22306.69876  41857.2132  22643.0594  22816.5776
## aar_f2011    -64358.98  22379.01911  22643.0594  45210.7304  23406.9880
## aar_f2012    -88195.78  22461.19628  22816.5776  23406.9880  47055.4187
## aar_f2013    -93332.22  22562.49160  23016.0483  23690.1311  24270.5328
## aar_f2014   -128032.51  22647.20878  23232.1454  24076.5421  24791.9383
## aar_f2015   -177893.27  22637.74268  23267.9132  24237.7165  25055.0255
## aar_f2016   -229170.12  22623.80635  23323.0788  24446.1520  25385.7301
## aar_f2017   -231919.09  22624.44448  23352.3686  24515.4258  25408.7607
## Total_ya_p   -134378.95    89.41919    277.8154    681.8928   1112.5721
## inc_k1       -48847.48   -46.78668   -117.7882    188.8338    193.4766
## inc_k5       -26724.41   110.78484    126.8286    397.1950    455.5137
## uni_k_mf     -23624.40  -129.42390   -212.3787   -468.5265   -572.7298
## uni_l_mf      79213.28   -45.36231   -237.3954   -324.3915   -491.9711
## Trade_pc_100K 145568.84   497.16540   1261.8579    987.3383    936.1196
##          aar_f2013    aar_f2014    aar_f2015    aar_f2016    aar_f2017
## (Intercept) -93332.21682 -128032.5143 -177893.2733 -229170.1243 -231919.0869
## aar_f2009    22562.49160  22647.20888  22637.7427  22623.8064  22624.4445
## aar_f2010    23016.04825  23232.1454  23267.9132  23323.0788  23352.3686
## aar_f2011    23690.13111  24076.5421  24237.7165  24446.1520  24515.4258
## aar_f2012    24270.53282  24791.9383  25055.0255  25385.7301  25408.7607
```

```

## aar_f2013      49220.90256    25428.8815    25755.4473    26135.5595    26169.5465
## aar_f2014      25428.88146    53475.4422    27156.8674    27482.0673    27045.3309
## aar_f2015      25755.44730    27156.8674    63394.1122    28309.5656    27655.2812
## aar_f2016      26135.55952    27482.0673    28309.5656    75087.4602    28071.1160
## aar_f2017      26169.54649    27045.3309    27655.2812    28071.1160    89424.5717
## Total_ya_p      1311.74280     1662.7240     2349.7551     3130.9906     3266.6554
## inc_k1          -23.25608      237.9932      438.1822      706.9105      723.9683
## inc_k5          419.80206      750.9501      927.6337     1166.2786     1178.1709
## uni_k_mf        -695.90501     -198.2867      136.4018     -110.1222     -816.2879
## uni_l_mf        -632.27758     -2195.0185    -3034.7846    -2540.7427    -1110.7783
## Trade_pc_100K   2510.69810     2684.4013     2764.2300     282.6406     1862.4720
##               Total_ya_p      inc_k1      inc_k5      uni_k_mf      uni_l_mf
## (Intercept)    -134378.94615  -48847.47803  -26724.4053  -23624.40438  79213.27980
## aar_f2009         89.41919     -46.78668      110.7848     -129.42390     -45.36231
## aar_f2010        277.81538     -117.78822      126.8286     -212.37867     -237.39541
## aar_f2011        681.89276      188.83384      397.1950     -468.52650     -324.39148
## aar_f2012       1112.57212      193.47663      455.5137     -572.72977     -491.97106
## aar_f2013       1311.74280     -23.25608      419.8021     -695.90501     -632.27758
## aar_f2014       1662.72401      237.99318      750.9501     -198.28673     -2195.01848
## aar_f2015       2349.75511      438.18220      927.6337      136.40176     -3034.78456
## aar_f2016       3130.99055      706.91052     1166.2786     -110.12216     -2540.74265
## aar_f2017       3266.65535      723.96826     1178.1709     -816.28793     -1110.77830
## Total_ya_p       2167.75020      426.37025      133.2185       51.21924     -614.02732
## inc_k1           426.37025      801.89764      496.4444      158.26504     -500.25996
## inc_k5           133.21845      496.44438      547.3448      104.53767     -690.28424
## uni_k_mf          51.21924      158.26504      104.5377     1515.96690     -2398.54359
## uni_l_mf         -614.02732     -500.25996     -690.2842     -2398.54359     5463.68941
## Trade_pc_100K   -1619.34164     -2293.03278     -115.1786     -2608.77275     651.94105
##               Trade_pc_100K
## (Intercept)     145568.8365
## aar_f2009         497.1654
## aar_f2010        1261.8579
## aar_f2011         987.3383
## aar_f2012         936.1196
## aar_f2013        2510.6981
## aar_f2014        2684.4013
## aar_f2015        2764.2300
## aar_f2016        282.6406
## aar_f2017        1862.4720
## Total_ya_p       -1619.3416
## inc_k1           -2293.0328
## inc_k5           -115.1786
## uni_k_mf         -2608.7728
## uni_l_mf          651.9410
## Trade_pc_100K    60897.1826

```

iv.

```

pm2 <- pm2 %>%
  add_residuals(lm1)

```

v.

```
pm2 <- pm2 %>%  
  mutate(aar_d = make_date(aar))
```

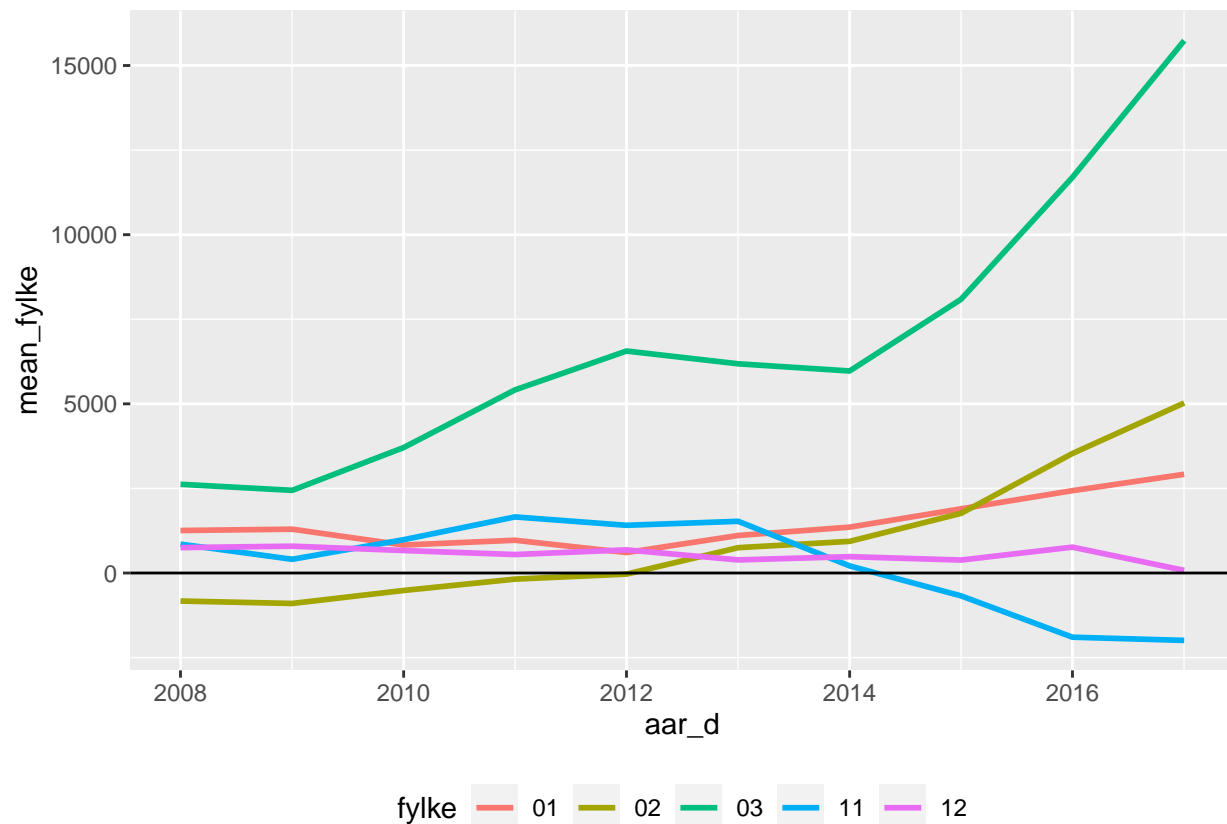
vi.

```
pm2 <- pm2 %>%  
  mutate(fylke = substr(knr, start = 1, stop = 2))
```

vii -x.

```
pm2 %>%  
  filter(fylke %in% c("01", "02", "03", "11", "12")) %>%  
  unnest(c(fylke)) %>%  
  group_by(fylke, aar_d) %>%  
  summarize(mean_fylke = mean(resid)  
            ) %>%  
  ggplot(aes(x = aar_d, y = mean_fylke, colour = fylke)) +  
  geom_line(lwd=1) +  
  theme(legend.position = "bottom")+  
  geom_hline(yintercept = 0, colour = "black")
```

'summarise()' has grouped output by 'fylke'. You can override using the '.groups' argument.



Dummy fylke og år

i & ii.

```
mod2 <- 'pm2 ~ aar_f*fnr + Total_ya_p + inc_k1 + inc_k5 + uni_k_mf + uni_l_mf + Trade_pc_100K'

lm2 <- lm(mod2, data = pm2)

summary(lm2)
```

```
##
## Call:
## lm(formula = mod2, data = pm2)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -8546  -1191      32    1198   8328
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -21200.688   2521.645  -8.407  < 2e-16 ***
## aar_f2009         94.009    744.240   0.126  0.899496
## aar_f2010        417.129    744.379   0.560  0.575290
```


## aar_f2011	1280.914	744.731	1.720	0.085597	.
## aar_f2012	1455.525	745.679	1.952	0.051088	.
## aar_f2013	2479.533	746.367	3.322	0.000910	***
## aar_f2014	2795.831	747.254	3.741	0.000188	***
## aar_f2015	3987.973	748.109	5.331	1.09e-07	***
## aar_f2016	5264.965	749.169	7.028	2.89e-12	***
## aar_f2017	6618.572	749.430	8.831	< 2e-16	***
## fnr02	-1482.789	702.970	-2.109	0.035045	*
## fnr03	3248.234	2190.443	1.483	0.138260	
## fnr04	-1049.219	774.264	-1.355	0.175537	
## fnr05	-1937.388	758.293	-2.555	0.010696	*
## fnr06	-2172.731	772.094	-2.814	0.004941	**
## fnr07	-737.995	1080.348	-0.683	0.494620	
## fnr08	-3213.279	878.620	-3.657	0.000262	***
## fnr09	-1219.813	913.691	-1.335	0.182020	
## fnr10	-281.375	852.265	-0.330	0.741323	
## fnr11	-565.360	771.927	-0.732	0.464012	
## fnr12	-903.071	742.464	-1.216	0.224012	
## fnr14	-3339.829	1182.013	-2.826	0.004768	**
## fnr15	-3619.198	715.832	-5.056	4.69e-07	***
## fnr16	-1093.217	759.677	-1.439	0.150296	
## fnr17	-2005.965	917.216	-2.187	0.028860	*
## fnr18	-1567.503	774.530	-2.024	0.043126	*
## fnr19	-2856.881	1326.142	-2.154	0.031341	*
## fnr20	-2656.315	1180.088	-2.251	0.024500	*
## Total_ya_p	511.787	36.100	14.177	< 2e-16	***
## inc_k1	-243.050	27.007	-9.000	< 2e-16	***
## inc_k5	251.645	22.916	10.981	< 2e-16	***
## uni_k_mf	178.253	28.157	6.331	3.02e-10	***
## uni_l_mf	732.442	42.235	17.342	< 2e-16	***
## Trade_pc_100K	1067.760	190.885	5.594	2.54e-08	***
## aar_f2009:fnr02	-40.505	978.026	-0.041	0.966969	
## aar_f2010:fnr02	792.694	978.020	0.811	0.417747	
## aar_f2011:fnr02	992.480	978.070	1.015	0.310359	
## aar_f2012:fnr02	1565.161	978.102	1.600	0.109716	
## aar_f2013:fnr02	1953.373	978.298	1.997	0.045996	*
## aar_f2014:fnr02	2019.269	978.649	2.063	0.039214	*
## aar_f2015:fnr02	2401.120	979.036	2.453	0.014273	*
## aar_f2016:fnr02	3656.344	979.067	3.735	0.000193	***
## aar_f2017:fnr02	4707.776	979.374	4.807	1.65e-06	***
## aar_f2009:fnr03	84.133	3068.211	0.027	0.978127	
## aar_f2010:fnr03	2004.378	3068.354	0.653	0.513677	
## aar_f2011:fnr03	3891.025	3068.768	1.268	0.204970	
## aar_f2012:fnr03	5674.403	3069.281	1.849	0.064642	.
## aar_f2013:fnr03	5108.375	3070.149	1.664	0.096297	.
## aar_f2014:fnr03	4938.603	3071.105	1.608	0.107979	
## aar_f2015:fnr03	6985.367	3073.112	2.273	0.023131	*
## aar_f2016:fnr03	10264.572	3074.072	3.339	0.000856	***
## aar_f2017:fnr03	13986.613	3075.071	4.548	5.74e-06	***
## aar_f2009:fnr04	-330.219	1089.318	-0.303	0.761813	
## aar_f2010:fnr04	-191.813	1089.355	-0.176	0.860250	
## aar_f2011:fnr04	-775.700	1089.399	-0.712	0.476523	
## aar_f2012:fnr04	-808.528	1089.510	-0.742	0.458115	
## aar_f2013:fnr04	-1206.685	1089.615	-1.107	0.268240	

```

## aar_f2014:fnr04 -1456.367 1089.708 -1.336 0.181550
## aar_f2015:fnr04 -1912.336 1089.754 -1.755 0.079446 .
## aar_f2016:fnr04 -2459.017 1089.893 -2.256 0.024169 *
## aar_f2017:fnr04 -3549.658 1089.920 -3.257 0.001146 **
## aar_f2009:fnr05 416.862 1069.758 0.390 0.696816
## aar_f2010:fnr05 655.342 1069.794 0.613 0.540221
## aar_f2011:fnr05 183.865 1069.834 0.172 0.863563
## aar_f2012:fnr05 820.104 1070.017 0.766 0.443507
## aar_f2013:fnr05 -198.536 1070.094 -0.186 0.852832
## aar_f2014:fnr05 -254.055 1070.253 -0.237 0.812388
## aar_f2015:fnr05 -1326.089 1070.254 -1.239 0.215480
## aar_f2016:fnr05 -2117.228 1070.338 -1.978 0.048059 *
## aar_f2017:fnr05 -2397.820 1070.176 -2.241 0.025165 *
## aar_f2009:fnr06 -163.759 1089.292 -0.150 0.880516
## aar_f2010:fnr06 189.332 1089.409 0.174 0.862046
## aar_f2011:fnr06 33.963 1089.394 0.031 0.975132
## aar_f2012:fnr06 800.976 1089.455 0.735 0.462302
## aar_f2013:fnr06 410.281 1089.375 0.377 0.706497
## aar_f2014:fnr06 571.152 1089.474 0.524 0.600167
## aar_f2015:fnr06 22.631 1089.626 0.021 0.983431
## aar_f2016:fnr06 -598.671 1089.701 -0.549 0.582801
## aar_f2017:fnr06 60.036 1089.704 0.055 0.956069
## aar_f2009:fnr07 134.353 1525.051 0.088 0.929808
## aar_f2010:fnr07 728.914 1525.112 0.478 0.632745
## aar_f2011:fnr07 275.017 1525.266 0.180 0.856930
## aar_f2012:fnr07 1047.940 1525.235 0.687 0.492122
## aar_f2013:fnr07 890.998 1525.236 0.584 0.559173
## aar_f2014:fnr07 582.123 1525.332 0.382 0.702772
## aar_f2015:fnr07 990.944 1525.354 0.650 0.515996
## aar_f2016:fnr07 447.813 1525.278 0.294 0.769099
## aar_f2017:fnr07 960.018 1525.236 0.629 0.529146
## aar_f2009:fnr08 329.317 1240.237 0.266 0.790631
## aar_f2010:fnr08 1281.636 1240.345 1.033 0.301597
## aar_f2011:fnr08 646.495 1240.336 0.521 0.602269
## aar_f2012:fnr08 1090.416 1240.413 0.879 0.379470
## aar_f2013:fnr08 575.599 1240.249 0.464 0.642628
## aar_f2014:fnr08 689.084 1240.251 0.556 0.578548
## aar_f2015:fnr08 -776.910 1240.290 -0.626 0.531130
## aar_f2016:fnr08 -1716.491 1240.468 -1.384 0.166595
## aar_f2017:fnr08 -2045.538 1240.415 -1.649 0.099294 .
## aar_f2009:fnr09 686.715 1288.922 0.533 0.594245
## aar_f2010:fnr09 986.486 1288.914 0.765 0.444149
## aar_f2011:fnr09 599.582 1288.944 0.465 0.641860
## aar_f2012:fnr09 1071.846 1289.011 0.832 0.405779
## aar_f2013:fnr09 64.585 1289.204 0.050 0.960050
## aar_f2014:fnr09 -186.541 1289.179 -0.145 0.884965
## aar_f2015:fnr09 -1242.730 1289.232 -0.964 0.335201
## aar_f2016:fnr09 -1987.219 1289.181 -1.541 0.123368
## aar_f2017:fnr09 -3223.036 1289.344 -2.500 0.012510 *
## aar_f2009:fnr10 231.288 1199.909 0.193 0.847172
## aar_f2010:fnr10 924.121 1199.916 0.770 0.441302
## aar_f2011:fnr10 168.648 1199.944 0.141 0.888243
## aar_f2012:fnr10 321.458 1200.216 0.268 0.788856
## aar_f2013:fnr10 -515.180 1200.200 -0.429 0.667793

```

## aar_f2014:fnr10	-674.319	1200.339	-0.562	0.574335	
## aar_f2015:fnr10	-1492.749	1200.502	-1.243	0.213856	
## aar_f2016:fnr10	-3090.918	1200.777	-2.574	0.010124	*
## aar_f2017:fnr10	-3807.142	1200.767	-3.171	0.001545	**
## aar_f2009:fnr11	-414.412	1069.772	-0.387	0.698515	
## aar_f2010:fnr11	642.468	1069.866	0.601	0.548235	
## aar_f2011:fnr11	1243.418	1070.024	1.162	0.245359	
## aar_f2012:fnr11	1467.212	1070.665	1.370	0.170728	
## aar_f2013:fnr11	1179.371	1071.062	1.101	0.270979	
## aar_f2014:fnr11	-183.391	1071.523	-0.171	0.864124	
## aar_f2015:fnr11	-1489.385	1072.451	-1.389	0.165063	
## aar_f2016:fnr11	-3274.743	1072.946	-3.052	0.002303	**
## aar_f2017:fnr11	-3863.610	1073.185	-3.600	0.000326	***
## aar_f2009:fnr12	21.853	1036.805	0.021	0.983186	
## aar_f2010:fnr12	381.898	1036.801	0.368	0.712658	
## aar_f2011:fnr12	165.379	1036.901	0.159	0.873297	
## aar_f2012:fnr12	669.171	1037.128	0.645	0.518864	
## aar_f2013:fnr12	-69.430	1037.183	-0.067	0.946636	
## aar_f2014:fnr12	-147.825	1037.277	-0.143	0.886690	
## aar_f2015:fnr12	-711.755	1037.476	-0.686	0.492767	
## aar_f2016:fnr12	-901.775	1037.688	-0.869	0.384941	
## aar_f2017:fnr12	-2046.447	1038.104	-1.971	0.048828	*
## aar_f2009:fnr14	-220.698	1663.985	-0.133	0.894498	
## aar_f2010:fnr14	536.844	1663.957	0.323	0.747009	
## aar_f2011:fnr14	1984.847	1664.012	1.193	0.233090	
## aar_f2012:fnr14	1739.551	1664.177	1.045	0.296018	
## aar_f2013:fnr14	208.353	1664.208	0.125	0.900381	
## aar_f2014:fnr14	253.302	1664.812	0.152	0.879084	
## aar_f2015:fnr14	-1695.187	1665.139	-1.018	0.308783	
## aar_f2016:fnr14	-1552.417	1665.259	-0.932	0.351330	
## aar_f2017:fnr14	-2074.192	1665.271	-1.246	0.213077	
## aar_f2009:fnr15	205.720	998.429	0.206	0.836779	
## aar_f2010:fnr15	548.008	998.671	0.549	0.583249	
## aar_f2011:fnr15	463.880	998.884	0.464	0.642414	
## aar_f2012:fnr15	463.860	999.265	0.464	0.642556	
## aar_f2013:fnr15	7.994	999.213	0.008	0.993617	
## aar_f2014:fnr15	-481.056	999.093	-0.481	0.630220	
## aar_f2015:fnr15	-587.449	999.385	-0.588	0.556727	
## aar_f2016:fnr15	-1872.887	999.582	-1.874	0.061126	.
## aar_f2017:fnr15	-2799.827	999.681	-2.801	0.005149	**
## aar_f2009:fnr16	-346.631	1069.772	-0.324	0.745955	
## aar_f2010:fnr16	-237.962	1069.934	-0.222	0.824020	
## aar_f2011:fnr16	-497.945	1069.952	-0.465	0.641705	
## aar_f2012:fnr16	380.682	1070.437	0.356	0.722154	
## aar_f2013:fnr16	-347.235	1070.757	-0.324	0.745754	
## aar_f2014:fnr16	-229.362	1070.812	-0.214	0.830418	
## aar_f2015:fnr16	-139.973	1070.880	-0.131	0.896019	
## aar_f2016:fnr16	-1074.143	1070.970	-1.003	0.316004	
## aar_f2017:fnr16	-2278.453	1070.923	-2.128	0.033499	*
## aar_f2009:fnr17	-288.412	1288.940	-0.224	0.822969	
## aar_f2010:fnr17	-422.338	1289.001	-0.328	0.743214	
## aar_f2011:fnr17	257.671	1289.086	0.200	0.841590	
## aar_f2012:fnr17	637.493	1289.624	0.494	0.621133	
## aar_f2013:fnr17	203.405	1289.762	0.158	0.874704	

```
## aar_f2014:fnr17    -61.073    1289.824   -0.047  0.962239
## aar_f2015:fnr17   -867.834    1289.740   -0.673  0.501107
## aar_f2016:fnr17  -1612.215    1290.487   -1.249  0.211703
## aar_f2017:fnr17  -2761.733    1290.527   -2.140  0.032479 *
## aar_f2009:fnr18   -148.285    1089.412   -0.136  0.891744
## aar_f2010:fnr18    402.939    1089.510    0.370  0.711545
## aar_f2011:fnr18    252.454    1089.674    0.232  0.816812
## aar_f2012:fnr18    482.679    1089.761    0.443  0.657871
## aar_f2013:fnr18    201.272    1090.026    0.185  0.853524
## aar_f2014:fnr18   -393.115    1090.258   -0.361  0.718459
## aar_f2015:fnr18   -439.127    1090.372   -0.403  0.687190
## aar_f2016:fnr18  -1361.291    1090.771   -1.248  0.212178
## aar_f2017:fnr18  -2661.041    1090.689   -2.440  0.014785 *
## aar_f2009:fnr19    453.061    1872.733    0.242  0.808864
## aar_f2010:fnr19    982.125    1872.779    0.524  0.600045
## aar_f2011:fnr19   -669.729    1872.850   -0.358  0.720682
## aar_f2012:fnr19    727.671    1872.902    0.389  0.697670
## aar_f2013:fnr19    278.261    1873.128    0.149  0.881921
## aar_f2014:fnr19   1688.165    1873.121    0.901  0.367563
## aar_f2015:fnr19    369.085    1873.412    0.197  0.843839
## aar_f2016:fnr19    906.286    1873.612    0.484  0.628646
## aar_f2017:fnr19   -716.410    1873.886   -0.382  0.702272
## aar_f2009:fnr20   -927.061    1664.164   -0.557  0.577542
## aar_f2010:fnr20   -547.207    1664.063   -0.329  0.742313
## aar_f2011:fnr20   -542.321    1664.293   -0.326  0.744568
## aar_f2012:fnr20   -378.342    1664.741   -0.227  0.820240
## aar_f2013:fnr20  -1110.163    1664.836   -0.667  0.504960
## aar_f2014:fnr20  -1563.827    1665.176   -0.939  0.347778
## aar_f2015:fnr20  -3266.760    1665.444   -1.961  0.049964 *
## aar_f2016:fnr20  -3169.910    1665.821   -1.903  0.057200 .
## aar_f2017:fnr20  -3922.387    1665.464   -2.355  0.018615 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2105 on 1944 degrees of freedom
## Multiple R-squared:  0.8953, Adjusted R-squared:  0.8848
## F-statistic: 85.21 on 195 and 1944 DF,  p-value: < 2.2e-16
```

iii.

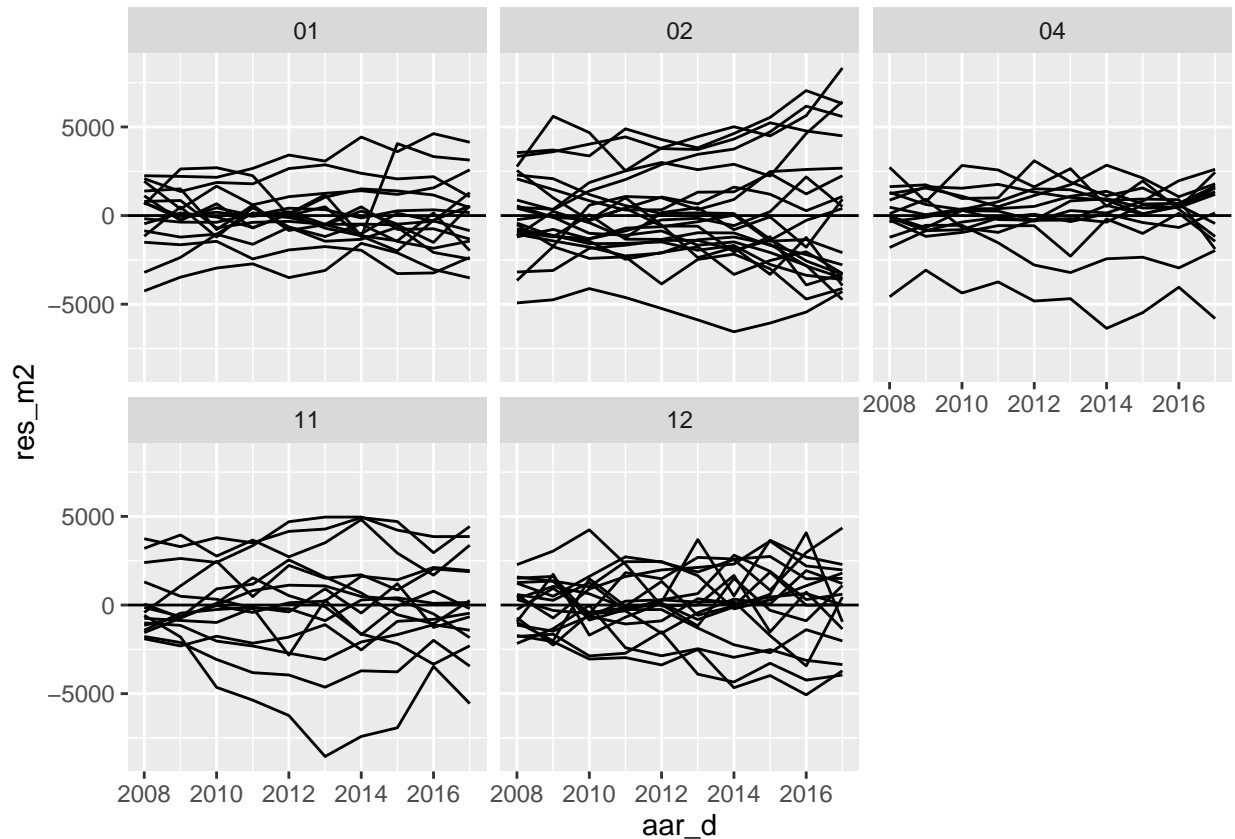
```
pm2 <- pm2 %>%
  mutate(res_m2 = resid(lm2))
```

iv.

Delplott:

```
pm2 %>% filter(fnr %in% c("01", "02", "04", "11", "12")) %>%
  ggplot(mapping = aes(x = aar_d, y = res_m2)) +
  geom_line(aes(group = knavn)) +
  scale_size_manual(values = c(seq(2.0, 0.5, by = -0.1))) +
```

```
geom_hline(yintercept = 0) +
theme(legend.position = 'bottom') +
  facet_wrap(~fylke)
```



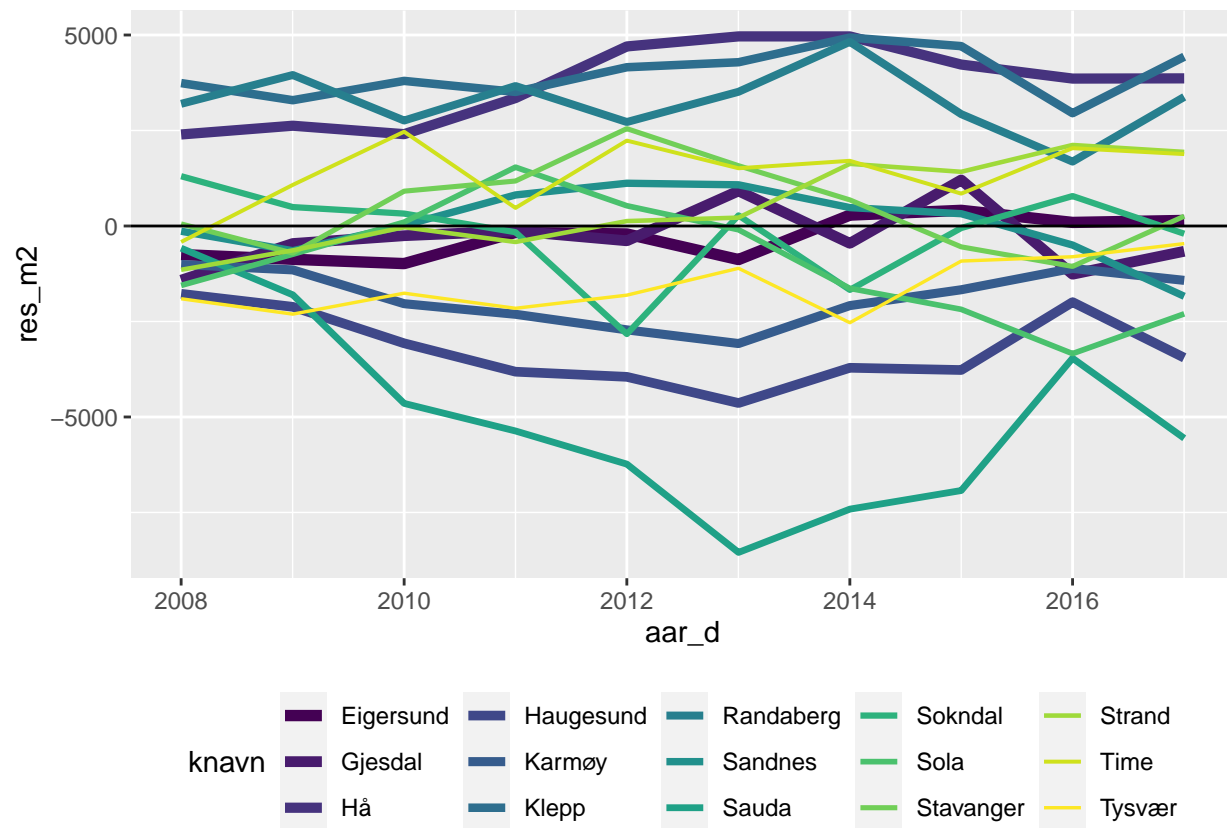
i & ii.

Kvaliteten på modellen er ikke helt optimal da den mangler noen variabler. Dette kan ha noe med heteroskedastisitet i modell at det er stor variasjon. Det er store residualer, spesielt i Rogaland.

Ut i fra grafene så ser man at variasjonen er stor. Dette indikerer et heteroskedastisitetsproblem, og dermed er det grunn til at det er utelatte viktige variabler (brudd på TS.3/TS'3)

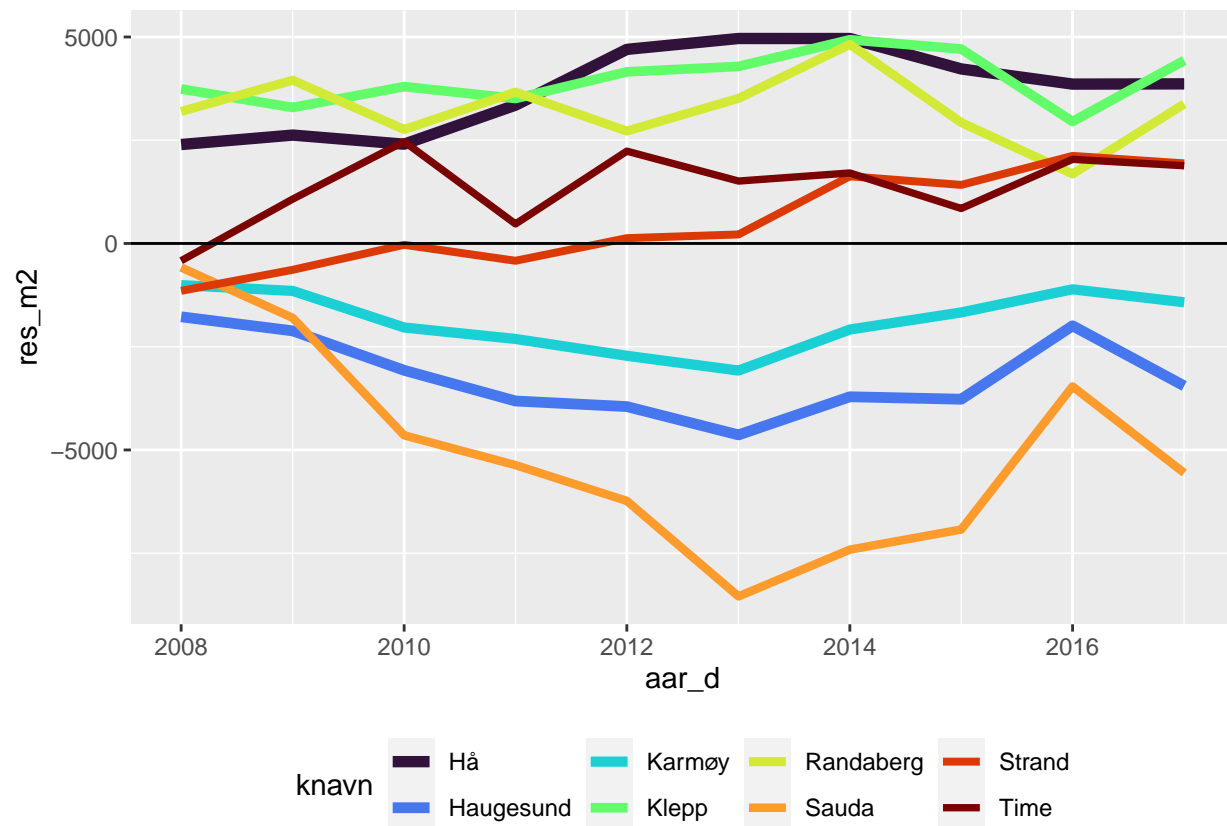
iii.

```
pm2 %>% filter(fnr %in% c("11")) %>%
ggplot(mapping = aes(x = aar_d, y = res_m2)) +
scale_color_viridis(discrete = TRUE, option = "D") +
geom_line(aes(group = knavn, colour = knavn, size = knavn)) +
scale_size_manual(values = c(seq(2.0, 0.5, by = -0.1))) +
geom_hline(yintercept = 0) +
theme(legend.position = 'bottom')
```



i.

```
pm2 %>% filter(knr %in% c("1119", "1120", "1127", "1121", "1130", "1135", "1106", "1149")) %>%
  ggplot(mapping = aes(x = aar_d, y = res_m2)) +
  scale_color_viridis(discrete = TRUE, option = "H") +
  geom_line(aes(group = knavn, colour = knavn, size = knavn)) +
  scale_size_manual(values = c(seq(2.0, 0.5, by = -0.1))) +
  geom_hline(yintercept = 0) +
  theme(legend.position = 'bottom')
```



ii.

Stavanger-kommunene overvurderes (Hå, Klepp og Randaberg).

Modell for hvert år

i.

```
pm2_n <- pm2 %>%
  group_by(aar) %>%
  select(pm2, fnr, knr, aar, aar_f, Menn_ya_p, Kvinner_ya_p, Total_ya_p, inc_k1, inc_k5, uni_k_mf, uni_k)
  nest()
```

```
pm2_n
```

```
## # A tibble: 10 x 2
## # Groups:   aar [10]
##   aar data
##   <dbl> <list>
## 1 2008 <tibble [214 x 12]>
## 2 2009 <tibble [214 x 12]>
```

```
## 3 2010 <tibble [214 x 12]>
## 4 2011 <tibble [214 x 12]>
## 5 2012 <tibble [214 x 12]>
## 6 2013 <tibble [214 x 12]>
## 7 2014 <tibble [214 x 12]>
## 8 2015 <tibble [214 x 12]>
## 9 2016 <tibble [214 x 12]>
## 10 2017 <tibble [214 x 12]>
```

```
pm2_n$data[[1]] %>%
head(n = 5)
```

```
## # A tibble: 5 x 12
##   pm2 fnr knr aar_f Menn_ya_p Kvinner_ya_p Total_ya_p inc_k1 inc_k5
##   <dbl> <chr> <chr> <chr>      <dbl>      <dbl>      <dbl> <dbl> <dbl>
## 1 13427 01 0101 2008      59.7      56.8      58.3  24.5  13.6
## 2 18299 01 0104 2008      60.7      58.7      59.7  22.8  16.2
## 3 14981 01 0105 2008      60.9      58.1      59.5  22.2  13.6
## 4 15671 01 0106 2008      59.8      57.8      58.8  21.8  16.2
## 5 18844 01 0111 2008      61.7      61.3      61.5  17.8  19
## # ... with 3 more variables: uni_k_mf <dbl>, uni_l_mf <dbl>,
## #   Trade_pc_100K <dbl>
```

```
dim(pm2_n)
```

```
## [1] 10 2
```

1.

```
kom_model <- function(a_df) {
  lm(pm2 ~ fnr + Total_ya_p + inc_k1 + inc_k5 + uni_k_mf + uni_l_mf + Trade_pc_100K, data = pm2)
}
```

```
pm2_n <- pm2_n %>%
  mutate(model = map(data, .f = kom_model))
```

```
kom_model(pm2_n$aar) %>%
  summary()
```

```
##
## Call:
## lm(formula = pm2 ~ fnr + Total_ya_p + inc_k1 + inc_k5 + uni_k_mf +
##   uni_l_mf + Trade_pc_100K, data = pm2)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -10648.8  -1602.8   -168.1   1474.5  14320.1
##
## Coefficients:
```



```
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)    6480.17   2872.86   2.256 0.024194 *
## fnr02          151.87    314.23   0.483 0.628913
## fnr03          7275.88   959.78   7.581 5.11e-14 ***
## fnr04         -2866.03    317.21  -9.035 < 2e-16 ***
## fnr05         -2728.80    305.31  -8.938 < 2e-16 ***
## fnr06         -2048.90    312.29  -6.561 6.70e-11 ***
## fnr07          -198.84    434.89  -0.457 0.647557
## fnr08         -3439.76    356.42  -9.651 < 2e-16 ***
## fnr09         -2211.31    367.84  -6.012 2.16e-09 ***
## fnr10         -1357.67    346.79  -3.915 9.33e-05 ***
## fnr11          -354.75    345.19  -1.028 0.304213
## fnr12         -1067.22    318.04  -3.356 0.000806 ***
## fnr14         -3685.59    483.41  -7.624 3.68e-14 ***
## fnr15         -3897.81    307.47 -12.677 < 2e-16 ***
## fnr16         -2039.39    304.44  -6.699 2.69e-11 ***
## fnr17         -3222.93    376.12  -8.569 < 2e-16 ***
## fnr18         -2229.33    316.67  -7.040 2.59e-12 ***
## fnr19         -2938.14    530.36  -5.540 3.40e-08 ***
## fnr20         -4283.12    477.15  -8.976 < 2e-16 ***
## Total_ya_p      136.67     42.46   3.219 0.001306 **
## inc_k1         -387.33     33.05 -11.720 < 2e-16 ***
## inc_k5           42.66     26.93   1.584 0.113318
## uni_k_mf        278.27     34.36   8.099 9.28e-16 ***
## uni_l_mf       1030.52     50.75  20.305 < 2e-16 ***
## Trade_pc_100K  1075.31    238.65   4.506 6.97e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2639 on 2115 degrees of freedom
## Multiple R-squared:  0.8209, Adjusted R-squared:  0.8189
## F-statistic: 403.9 on 24 and 2115 DF, p-value: < 2.2e-16
```

i.

```
pm2_n %>%
  filter(aar%in% c("2008")) %>%
  .$model %>%
  map_df(glance) %>%
  print()
```

```
## # A tibble: 1 x 12
##   r.squared adj.r.squared sigma statistic p.value    df logLik   AIC   BIC
##   <dbl>      <dbl> <dbl>    <dbl>   <dbl> <dbl> <dbl> <dbl> <dbl>
## 1    0.821        0.819 2639.    404.     0    24 -19883. 39818. 39965.
## # ... with 3 more variables: deviance <dbl>, df.residual <int>, nobs <int>
```

```
mod_sum <- pm2_n %>%
  filter(aar %in% c("2008", "2009", "2010", "2011", "2012", "2013", "2014", "2015", "2016", "2017")) %>%
  mutate(mod_summary = map(.x = model, .f = glance)) %>%
  unnest(mod_summary) %>%
  print()
```

```
## # A tibble: 10 x 15
## # Groups:   aar [10]
##       aar data      model r.squared adj.r.squared sigma statistic p.value    df
##   <dbl> <list>    <lis>    <dbl>         <dbl> <dbl>    <dbl>    <dbl> <dbl>
## 1  2008 <tibble [2~ <lm>      0.821         0.819 2639.    404.      0     24
## 2  2009 <tibble [2~ <lm>      0.821         0.819 2639.    404.      0     24
## 3  2010 <tibble [2~ <lm>      0.821         0.819 2639.    404.      0     24
## 4  2011 <tibble [2~ <lm>      0.821         0.819 2639.    404.      0     24
## 5  2012 <tibble [2~ <lm>      0.821         0.819 2639.    404.      0     24
## 6  2013 <tibble [2~ <lm>      0.821         0.819 2639.    404.      0     24
## 7  2014 <tibble [2~ <lm>      0.821         0.819 2639.    404.      0     24
## 8  2015 <tibble [2~ <lm>      0.821         0.819 2639.    404.      0     24
## 9  2016 <tibble [2~ <lm>      0.821         0.819 2639.    404.      0     24
## 10 2017 <tibble [2~ <lm>      0.821         0.819 2639.    404.      0     24
## # ... with 6 more variables: logLik <dbl>, AIC <dbl>, BIC <dbl>,
## #   deviance <dbl>, df.residual <int>, nobs <int>
```