

Modeller

```
# Jeg leser fra data mappen. Ser at det er en liten forskjell i størrelsen
# på de to csv filene
pm2 <- read_csv("data/pm2.csv", show_col_types = FALSE)
```

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

```
head(pm2)
```

```
## # A tibble: 6 x 18
##   knr      aar knavn   pm2 Menn_ya_p Kvinner_ya_p Total_ya_p inc_k1 inc_k5
##   <chr> <dbl> <chr>  <dbl>    <dbl>      <dbl>      <dbl> <dbl> <dbl>
## 1 0101   2008 Halden 13427    59.7        56.8        58.3  24.5  13.6
## 2 0101   2009 Halden 13095    59.8        57.0        58.4  24.4  14.1
## 3 0101   2010 Halden 13832    59.6        57.1        58.3  23.9  13.7
## 4 0101   2011 Halden 14915    59.8        57.2        58.5   24    14
## 5 0101   2012 Halden 15473    59.5        57.0        58.2  23.9   14
## 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>
```

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

```
## # A tibble: 2,140 x 18
##   knr      aar knavn   pm2 Menn_ya_p Kvinner_ya_p Total_ya_p inc_k1 inc_k5
##   <chr> <dbl> <chr>  <dbl>    <dbl>      <dbl>      <dbl> <dbl> <dbl>
## 1 0101   2008 Halden 13427    59.7        56.8        58.3  24.5  13.6
## 2 0101   2009 Halden 13095    59.8        57.0        58.4  24.4  14.1
## 3 0101   2010 Halden 13832    59.6        57.1        58.3  23.9  13.7
## 4 0101   2011 Halden 14915    59.8        57.2        58.5   24    14
## 5 0101   2012 Halden 15473    59.5        57.0        58.2  23.9   14
## 6 0101   2013 Halden 15461    59.0        56.7        57.9  24.1  13.4
## 7 0101   2014 Halden 17164    58.8        56.7        57.7  23.9  13.5
## 8 0101   2015 Halden 17427    58.7        56.8        57.8   24    13.7
## 9 0101   2016 Halden 18941    58.7        56.6        57.7   24    13.8
## 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>
```

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

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

```
## # A tibble: 4 x 19
##   knr      aar knavn    pm2 Menn_ya_p Kvinner_ya_p Total_ya_p inc_k1 inc_k5
##   <chr> <dbl> <chr>  <dbl>      <dbl>      <dbl>      <dbl> <dbl> <dbl>
## 1 0101   2008 Halden 13427      59.7       56.8       58.3  24.5  13.6
## 2 0101   2009 Halden 13095      59.8       57.0       58.4  24.4  14.1
## 3 0101   2010 Halden 13832      59.6       57.1       58.3  23.9  13.7
## 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'
```

```
lm1 = lm(mod1, data = pm2)
```

```
summary(lm1)
```

```
##
## Call:
## lm(formula = mod1, data = 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

```
pm2 %>%
  add_residuals(lm1)

## # A tibble: 2,140 x 20
##   knr      aar knavn      pm2 Menn_ya_p Kvinner_ya_p Total_ya_p inc_k1 inc_k5
##   <chr> <dbl> <chr> <dbl>      <dbl>      <dbl>      <dbl> <dbl> <dbl>
## 1 0101   2008 Halden 13427      59.7      56.8      58.3  24.5  13.6
## 2 0101   2009 Halden 13095      59.8      57.0      58.4  24.4  14.1
## 3 0101   2010 Halden 13832      59.6      57.1      58.3  23.9  13.7
## 4 0101   2011 Halden 14915      59.8      57.2      58.5  24    14
## 5 0101   2012 Halden 15473      59.5      57.0      58.2  23.9  14
## 6 0101   2013 Halden 15461      59.0      56.7      57.9  24.1  13.4
## 7 0101   2014 Halden 17164      58.8      56.7      57.7  23.9  13.5
## 8 0101   2015 Halden 17427      58.7      56.8      57.8  24    13.7
## 9 0101   2016 Halden 18941      58.7      56.6      57.7  24    13.8
## 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>
```

Residualene fra linær modell i datasettet pm2

```
pm2 %>%
  add_residuals(lm1)

## # A tibble: 2,140 x 20
##   knr      aar knavn      pm2 Menn_ya_p Kvinner_ya_p Total_ya_p inc_k1 inc_k5
##   <chr> <dbl> <chr> <dbl>      <dbl>      <dbl>      <dbl> <dbl> <dbl>
## 1 0101   2008 Halden 13427      59.7      56.8      58.3  24.5  13.6
## 2 0101   2009 Halden 13095      59.8      57.0      58.4  24.4  14.1
## 3 0101   2010 Halden 13832      59.6      57.1      58.3  23.9  13.7
## 4 0101   2011 Halden 14915      59.8      57.2      58.5  24    14
## 5 0101   2012 Halden 15473      59.5      57.0      58.2  23.9  14
## 6 0101   2013 Halden 15461      59.0      56.7      57.9  24.1  13.4
## 7 0101   2014 Halden 17164      58.8      56.7      57.7  23.9  13.5
## 8 0101   2015 Halden 17427      58.7      56.8      57.8  24    13.7
## 9 0101   2016 Halden 18941      58.7      56.6      57.7  24    13.8
## 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 19
##   knr      aar knavn      pm2 Menn_ya_p Kvinner_ya_p Total_ya_p inc_k1 inc_k5
##   <chr> <dbl> <chr> <dbl>      <dbl>      <dbl>      <dbl> <dbl> <dbl>
## 1 0101   2008 Halden 13427      59.7      56.8      58.3  24.5  13.6
```

```
## 2 0101    2009 Halden 13095      59.8      57.0      58.4    24.4    14.1
## 3 0101    2010 Halden 13832      59.6      57.1      58.3    23.9    13.7
## 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>
```

Forklaring

i

I år 2009 øker pm2 104, i 2010 med 908, ..., i år 2017 øker pm2 med 5146. Året 2009 er ikke signifikant, men koeffisientene er signifikante på 0.1% signifikansnivå fra 2010-2017. Her forekommer økning i koeffisientene fra et år til neste.

Heteroskedastisitet

i.

```
bptest(lm1)
```

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

```
# smart!
```

```
library(gvlma)
gvlma(lm1)
```

```
##
## Call:
## lm(formula = mod1, data = pm2)
##
## Coefficients:
## (Intercept)      aar_f2009      aar_f2010      aar_f2011      aar_f2012
## -20400.74         104.15         908.13         1663.93         2240.48
##      aar_f2013      aar_f2014      aar_f2015      aar_f2016      aar_f2017
##  2869.30         2863.22         3525.22         4274.99         5146.33
##      Total_ya_p      inc_k1      inc_k5      uni_k_mf      uni_l_mf
##    582.44        -376.99         194.35        -82.02        1206.86
## Trade_pc_100K
##      871.99
##
##
## ASSESSMENT OF THE LINEAR MODEL ASSUMPTIONS
## USING THE GLOBAL TEST ON 4 DEGREES-OF-FREEDOM:
## Level of Significance = 0.05
##
## Call:
## gvlma(x = lm1)
##
##
## Value p-value Decision
## Global Stat 733.35 0.000e+00 Assumptions NOT satisfied!
## Skewness 48.82 2.804e-12 Assumptions NOT satisfied!
```

```
## Kurtosis          538.05 0.000e+00 Assumptions NOT satisfied!
## Link Function      96.62 0.000e+00 Assumptions NOT satisfied!
## Heteroscedasticity 49.86 1.652e-12 Assumptions NOT satisfied!
```

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  22647.2088  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 %>%  
  # må paste på "-01-01" for at date skal virke  
  # make_date ser ut til å lage et datetime objekt  
  # Vi trenger ikke tidspunkt ;-)  
  mutate(aar_d = date(paste0(aar, "-01-01")))
```

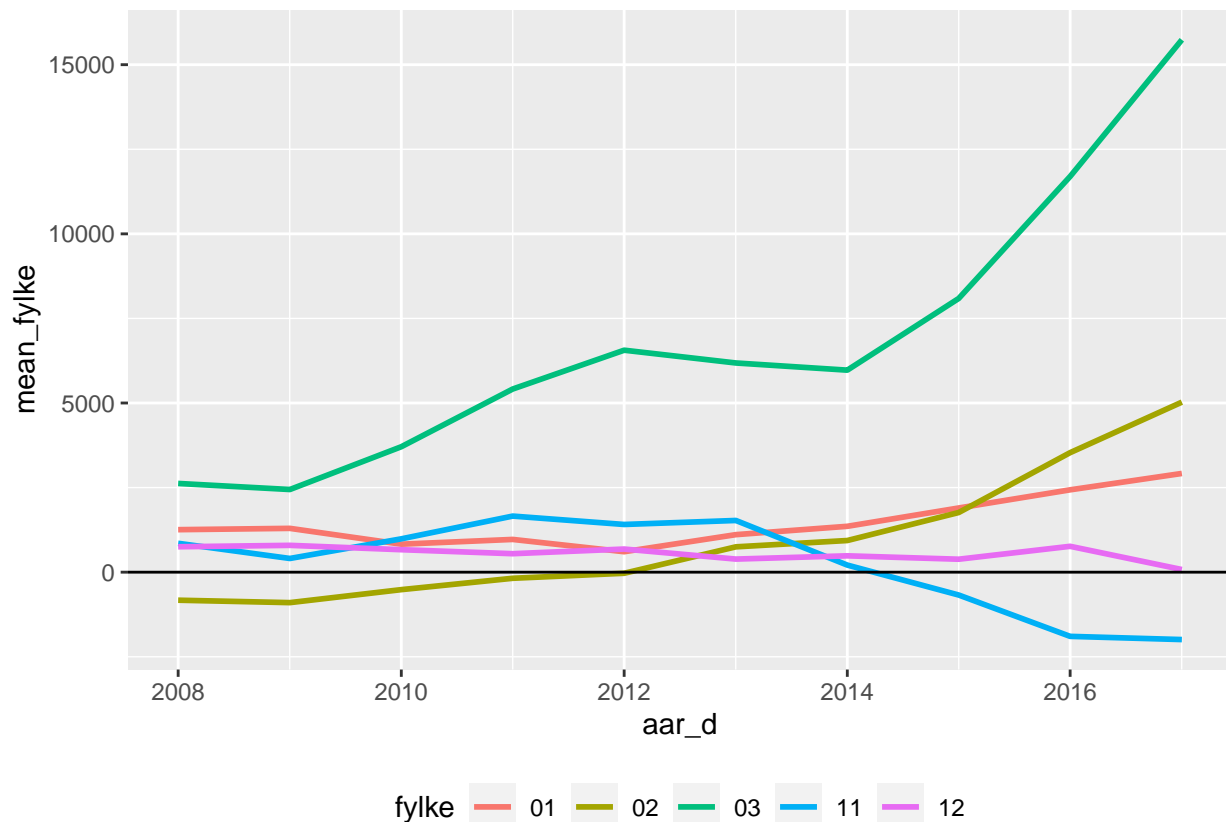
vi.

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

vii og viii.

```
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 og ii.

Innfører en dummy for hvert fylke hvert år. (Husk * gir interaksjonsvariabler automatisk i Rs formula).
Bruk interaksjon mellom fnr og aar_f istedenfor aar_f. La modell 2 ellers være lik modell 1.

```
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
```

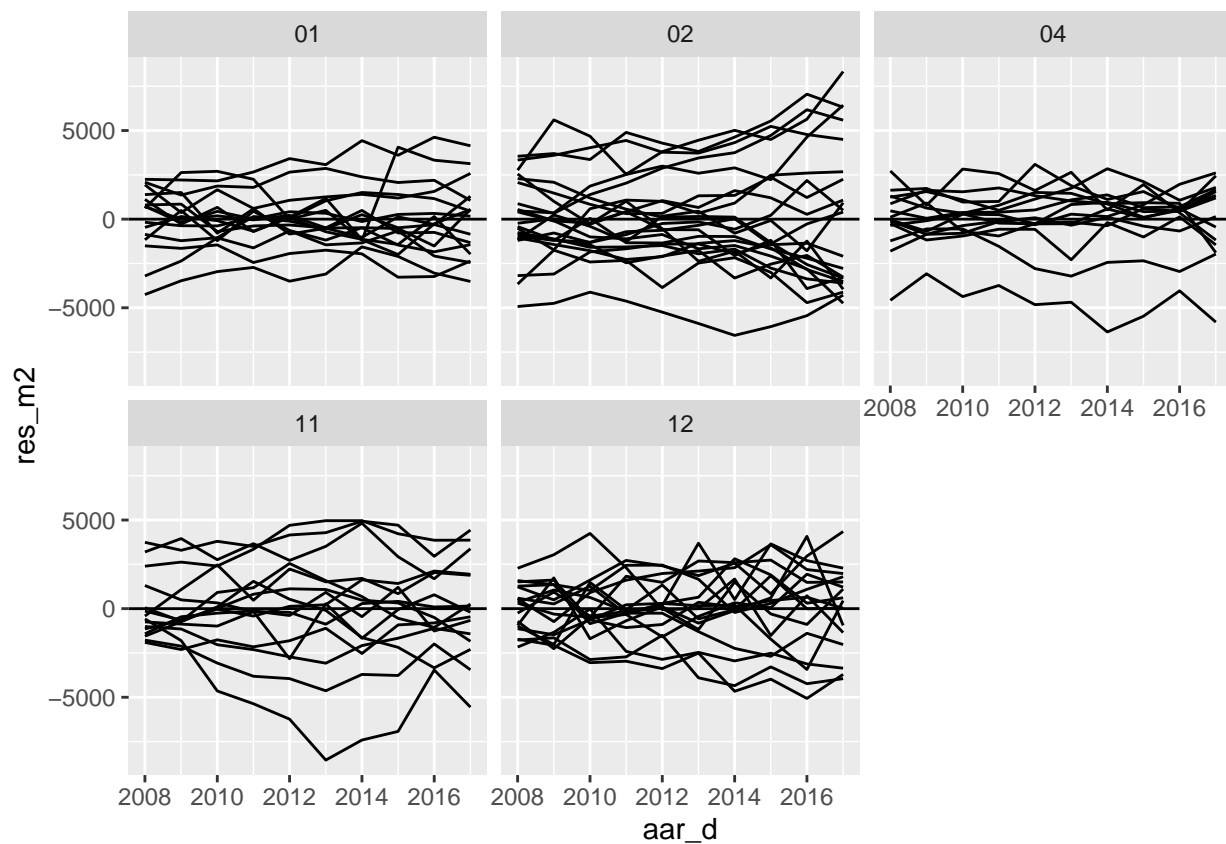
iii.

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

iv.

Delploott

```
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)
```

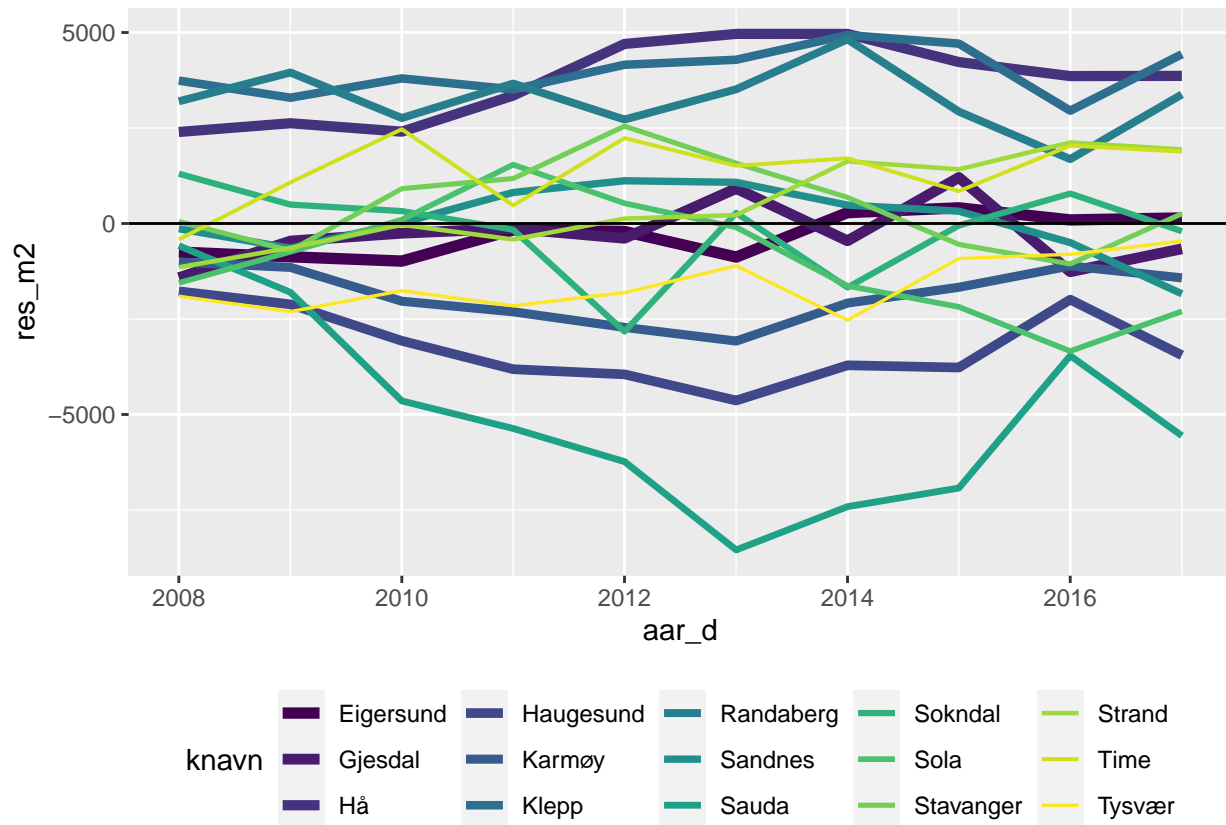


i, ii.

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.

iii.

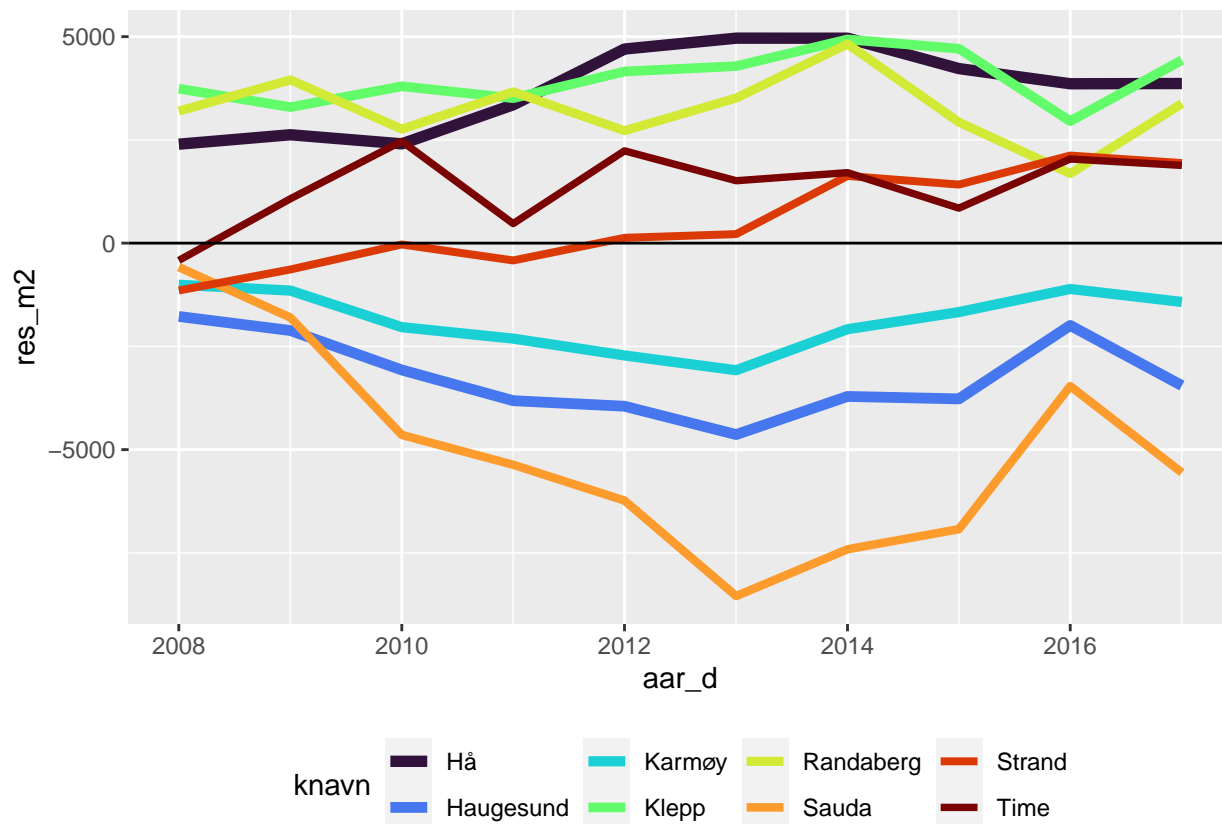
```
pm2 %>% filter(fnr %in% c("11")) %>%
  ggplot(mapping = aes(x = aar_d, y = res_m2)) +
  scale_color_viridis(discrete = TRUE, option = "D") +
  geom_line(aes(group = knavn, colour = knavn, size = knavn)) +
  scale_size_manual(values = c(seq(2.0, 0.5, by = -0.1))) +
  geom_hline(yintercept = 0) +
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



Det er ikke stor nok fargeskala for alle kommunene.

#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.

De som ligger nærmere Stavanger overvurderes.

Modell for hvert år

```
pm2_n <- pm2 %>%
  # tar med aar_d. Velger først variablene
  select(pm2, fnr, knr, aar, aar_f, aar_d, Menn_ya_p, Kvinner_ya_p, Total_ya_p, inc_k1, inc_k5, uni_k_m)
  group_by(aar) %>%
  nest()
```

pm2_n

```
## # A tibble: 10 x 2
## # Groups:   aar [10]
##   aar data
##   <dbl> <list>
## 1 2008 <tibble [214 x 13]>
## 2 2009 <tibble [214 x 13]>
## 3 2010 <tibble [214 x 13]>
## 4 2011 <tibble [214 x 13]>
## 5 2012 <tibble [214 x 13]>
## 6 2013 <tibble [214 x 13]>
## 7 2014 <tibble [214 x 13]>
## 8 2015 <tibble [214 x 13]>
```

```

## 9 2016 <tibble [214 x 13]>
## 10 2017 <tibble [214 x 13]>

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

## # A tibble: 5 x 13
##   pm2 fnr knr aar_f aar_d Menn_ya_p Kvinner_ya_p Total_ya_p inc_k1
##   <dbl> <chr> <chr> <chr> <date> <dbl> <dbl> <dbl> <dbl>
## 1 13427 01 0101 2008 2008-01-01 59.7 56.8 58.3 24.5
## 2 18299 01 0104 2008 2008-01-01 60.7 58.7 59.7 22.8
## 3 14981 01 0105 2008 2008-01-01 60.9 58.1 59.5 22.2
## 4 15671 01 0106 2008 2008-01-01 59.8 57.8 58.8 21.8
## 5 18844 01 0111 2008 2008-01-01 61.7 61.3 61.5 17.8
## # ... with 4 more variables: inc_k5 <dbl>, uni_k_mf <dbl>, uni_l_mf <dbl>,
## # Trade_pc_100K <dbl>

dim(pm2_n)

## [1] 10 2

# data må være lik a_df som er argumentet
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 = a_df)
}

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

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

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 0.873 0.857 1701. 54.2 1.19e-71 24 -1882. 3817. 3904.
## # ... 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 [~ <lm> 0.873 0.857 1701. 54.2 1.19e-71 24
## 2 2009 <tibble [~ <lm> 0.886 0.871 1614. 61.2 5.63e-76 24
## 3 2010 <tibble [~ <lm> 0.888 0.874 1743. 62.4 1.13e-76 24
## 4 2011 <tibble [~ <lm> 0.883 0.868 1925. 59.4 6.50e-75 24

```

```
## 5 2012 <tibble [~ <lm>      0.891      0.877 1953.      64.2 1.06e-77      24
## 6 2013 <tibble [~ <lm>      0.895      0.881 2026.      67.0 3.03e-79      24
## 7 2014 <tibble [~ <lm>      0.884      0.869 2149.      60.1 2.30e-75      24
## 8 2015 <tibble [~ <lm>      0.879      0.863 2361.      57.1 1.57e-73      24
## 9 2016 <tibble [~ <lm>      0.883      0.869 2467.      59.7 4.19e-75      24
## 10 2017 <tibble [~ <lm>      0.895      0.882 2614.      67.0 2.84e-79      24
## # ... with 6 more variables: logLik <dbl>, AIC <dbl>, BIC <dbl>,
## #   deviance <dbl>, df.residual <int>, nobs <int>

# Plukker ut modeeellen
coef_df <- mod_sum$model %>%
  # 1 henter ut koeffisientene, 2 residualene etc.
  map_df(1) %>%
  # trenger ikke endre til tibble, bare si at vi ønsker en tibble
  tibble()
```

Den siste delen mangler

Her er min kode for avslutningen.

i. Lag en ny variabel av type date i `coef_df` som angir år.

```
coef_df <- coef_df %>%

mutate(

aar = ymd(paste(2008:2017, "-01-01", sep = ""))

) %>%

select(aar, everything())
```

i. Pivot_longer `coef_df` til `coef_df_long`.

```
coef_df_long <- coef_df %>%

pivot_longer(

cols = `(Intercept)`:`Trade_pc_100K`,

names_to = "variables",

values_to = "coef"

)
```

i. Lag så et plot av fylke-faktorvariablenes koeffisienter for fylkene “fnr02”, “fnr03”, “fnr04”, “fnr10”, “fnr11”, “fnr12”, “fnr14” fra år 2008 til 2017.

ii. Hva sier plot-et oss om prisutviklingen i disse fylkene?

iii. Hva skjedde i 2014?

Her er kode for å lage plottet:

```
coef_df_long %>%

select(aar, variables, coef) %>%
```



```

filter(
  variables %in% c("fnr02", "fnr03", "fnr04", "fnr10", "fnr11", "fnr12", "fnr14")
) %>%

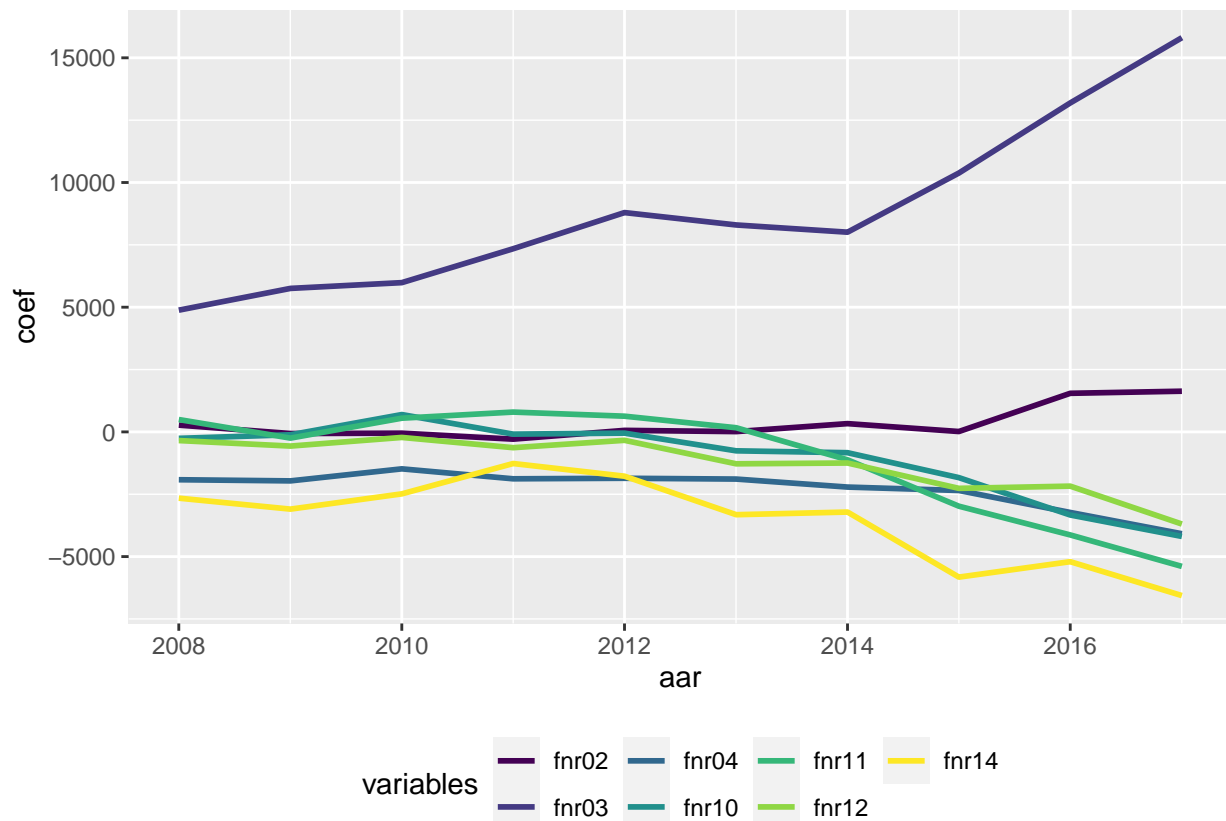
ggplot(mapping = aes(x = aar, y = coef, colour = variables)) +

scale_color_viridis(discrete = TRUE, option = "D") +

geom_line(aes(group = variables), lwd = 1) +

theme(legend.position = 'bottom')

```



i. Lag et plot tilsvarende det ovenfor for fnr, men nå for variablene Total_ya_p, inc_k1, inc_k5, uni_k_mf, uni_l_mf og Trade_pc_100K. (Plottet er gjengitt nedenfor, dere skal gjenskape det vha ggplot)

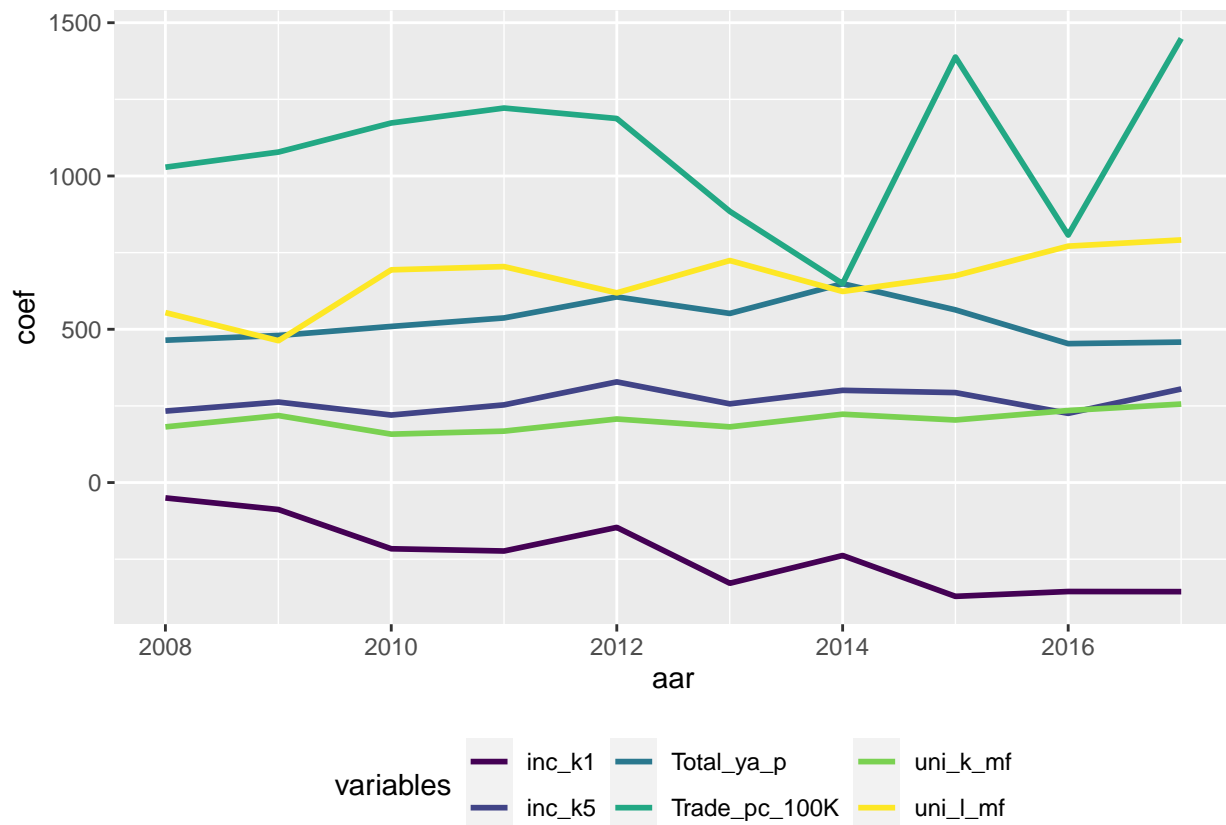
ii. Diskuter om koeffisientene ser ut til å være stabile over tid.

```

coef_df_long %>%
select(aar, variables, coef) %>%
filter(variables %in% c("Total_ya_p", "inc_k1", "inc_k5",
"uni_k_mf", "uni_l_mf", "Trade_pc_100K"))
) %>%
ggplot(
  mapping = aes(x = aar, y = coef, colour = variables)
) +

```

```
scale_color_viridis(discrete = TRUE, option = "D") +
geom_line(aes(group = variables), lwd = 1) +
theme(legend.position = 'bottom')
```



#siste