To complete the analysis on the significance of the sector on the salary for different occupational groups in Sweden I will in this post examine the correlation between salary and sector using statistics for education.

The F-value from the Anova table is used as the single value to discriminate how much the region and salary correlates. For exploratory analysis, the Anova value seems good enough.

First, define libraries and functions.

```
library (tidyverse)
## -- Attaching packages -------
tidyverse 1.3.0 --
                   v purrr 0.3.3
## v ggplot2 3.2.1
## v tibble 2.1.3
                   v dplyr 0.8.3
## v tidyr 1.0.2 v stringr 1.4.0
## v readr 1.3.1
                   v forcats 0.4.0
## -- Conflicts ---------------
tidyverse conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
library (broom)
library (car)
## Loading required package: carData
## Attaching package: 'car'
## The following object is masked from 'package:dplyr':
##
##
      recode
## The following object is masked from 'package:purrr':
##
##
      some
library (sjPlot)
## Registered S3 methods overwritten by 'lme4':
##
   method
## cooks.distance.influence.merMod car
##
   influence.merMod
                                 car
## dfbeta.influence.merMod
                                car
## dfbetas.influence.merMod
                                 car
## Learn more about sjPlot with 'browseVignettes("sjPlot")'.
readfile <- function (file1) {read_csv (file1, col_types = cols(), locale =</pre>
readr::locale (encoding = "latin1"), na = c("..", "NA")) %>%
 gather (starts with("19"), starts with("20"), key = "year", value = salary)
응>응
 drop_na() %>%
 mutate (year n = parse number (year))
```

The data table is downloaded from Statistics Sweden. It is saved as a comma-delimited file without heading, 000000CY.csv, http://www.statistikdatabasen.scb.se/pxweb/en/ssd/.

I have renamed the file to 000000CY_sector.csv because the filename 000000CY.csv was used in a previous post.

The table: Average basic salary, monthly salary and women's salary as a percentage of men's salary by sector, occupational group (SSYK 2012), sex and educational level (SUN). Year 2014 – 2018 Monthly salary 1-3 public sector 4-5 private sector

In the plot and tables, you can also find information on how the increase in salaries per year for each occupational group is affected when the interactions are taken into account.

```
tb <- readfile("000000CY_sector.csv") %>%
  mutate(edulevel = `level of education`)

numedulevel <- read.csv("edulevel.csv")

numedulevel %>%
  knitr::kable(
  booktabs = TRUE,
  caption = 'Initial approach, length of education')
```

Table 1: Initial approach, length of education

```
primary and secondary education 9-10 years (ISCED97 2)
upper secondary education, 2 years or less (ISCED97 3C)
11
upper secondary education 3 years (ISCED97 3A)
12
post-secondary education, less than 3 years (ISCED97 4+5B)
14
post-secondary education 3 years or more (ISCED97 5A)
15
post-graduate education (ISCED97 6)
19
no information about level of educational attainment
NA
```

```
no information about level of educational attainment
tbnum <- tb %>%
  right join(numedulevel, by = c("level of education" = "level.of.education"))
응>응
  filter(!is.na(eduyears)) %>%
 mutate(eduyears = factor(eduyears))
## Warning: Column `level of education`/`level.of.education` joining character
## vector and factor, coercing into character vector
summary table = vector()
anova table = vector()
for (i in unique(tbnum$`occuptional (SSYK 2012)`)){
 temp <- filter(tbnum, `occuptional (SSYK 2012)` == i)</pre>
  if (dim(temp)[1] > 90){
    model < -lm(log(salary) \sim edulevel + sex + year n + sector, data = temp)
    summary_table <- rbind (summary_table, mutate (tidy (summary (model)), ssyk</pre>
= i, interaction = "none"))
    anova table <- rbind (anova table, mutate (tidy (Anova (model, type = 2)),
ssyk = i, interaction = "none"))
    model <- lm(log(salary) ~ edulevel * sector + sex + year n, data = temp)</pre>
    summary table <- rbind (summary table, mutate (tidy (summary (model)), ssyk
= i, interaction = "sector and edulevel"))
    anova table <- rbind (anova table, mutate (tidy (Anova (model, type = 2)),
ssyk = i, interaction = "sector and edulevel"))
    model <- lm(log(salary) ~ edulevel + sector * sex + year n, data = temp)</pre>
```

```
summary table <- rbind (summary table, mutate (tidy (summary (model)), ssyk
= i, interaction = "sector and sex"))
    anova table <- rbind (anova table, mutate (tidy (Anova (model, type = 2)),
ssyk = i, interaction = "sector and sex"))
    model <- lm(log(salary) ~ edulevel + year n * sector + sex, data = temp)</pre>
    summary table <- rbind (summary table, mutate (tidy (summary (model)), ssyk
= i, interaction = "sector and year"))
    anova table <- rbind (anova table, mutate (tidy (Anova (model, type = 2)),
ssyk = i, interaction = "sector and year"))
   model <-lm(log(salary) \sim edulevel * sector * sex * year n, data = temp)
    summary table <- rbind (summary_table, mutate (tidy (summary (model)), ssyk</pre>
= i, interaction = "sector, year, edulevel and sex"))
    anova table <- rbind (anova table, mutate (tidy (Anova (model, type = 2)),
ssyk = i, interaction = "sector, year, edulevel and sex"))
 }
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anova_table <- anova table %>% rowwise() %>% mutate(contcol = str_count(term,
summary table <- summary table %>% rowwise() %>% mutate(contcol =
str count(term, ":"))
merge(summary_table, anova_table, by = c("ssyk", "interaction"), all = TRUE) \$>\$
 filter (term.x == "year n") %>%
 filter (term.y == "sector") %>%
 filter (interaction == "none") %>%
 mutate (estimate = (exp(estimate) - 1) * 100) %>%
  ggplot () +
    geom point (mapping = aes(x = estimate, y = statistic.y, colour = estimate)
interaction)) +
     x = "Increase in salaries (% / year)",
     y = "F-value for sector"
```

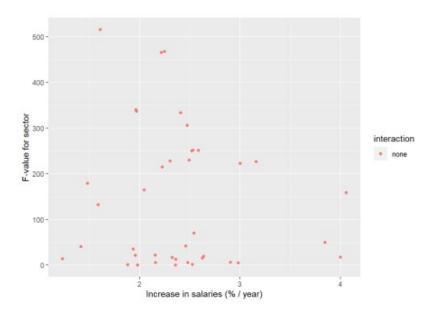


Figure 1: The significance of the sector on the salary in Sweden, a comparison between different occupational groups, Year 2014 - 2018

```
merge(summary_table, anova_table, by = c("ssyk", "interaction"), all = TRUE) %>%
  filter (term.x == "year_n") %>%
  filter (contcol.y > 0) %>%
  # only look at the interactions between all four variables in the case with
interaction sector, year, edulevel and sex
  filter (!(contcol.y < 3 & interaction == "sector, year, edulevel and sex"))
%>%

mutate (estimate = (exp(estimate) - 1) * 100) %>%
  ggplot () +
  geom_point (mapping = aes(x = estimate, y = statistic.y, colour =
interaction)) +
  labs(
    x = "Increase in salaries (% / year)",
    y = "F-value for interaction"
)
```

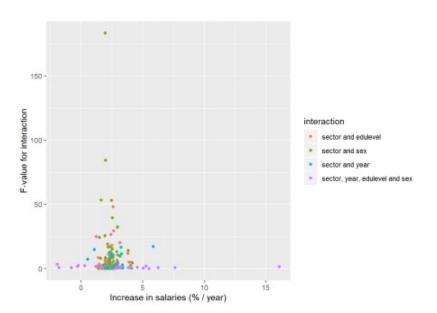


Figure 2: The significance of the interaction between sector, edulevel, year and sex on the salary in Sweden, a comparison between different occupational groups, Year 2014 – 2018

The tables with all occupational groups sorted by F-value in descending order.

```
merge(summary_table, anova_table, c("ssyk", "interaction"), all = TRUE) %>%
  filter (term.x == "year_n") %>%
  filter (term.y == "sector") %>%
  filter (interaction == "none") %>%
  mutate (estimate = (exp(estimate) - 1) * 100) %>%
  select (ssyk, estimate, statistic.y, interaction) %>%
  rename (`F-value` = statistic.y) %>%
  rename (`Increase in salary` = estimate) %>%
  arrange (desc (`F-value`)) %>%
  knitr::kable(
    booktabs = TRUE,
    caption = 'Correlation for F-value (sector) and the yearly increase in salaries')
```

Table 2: Correlation for F-value (sector) and the yearly increase in salaries

ssyk	Increase in F-value interaction salary
242 Organisation analysts, policy administrators and human resource specialists	1.609869 515.5748558 none
819 Process control technicians	2.250895 467.4732796 none
251 ICT architects, systems analysts and test managers	2.217446 465.2253589 none
331 Financial and accounting associate professionals	1.964825 340.2901702 none
962 Newspaper distributors, janitors and other service workers	1.971413 336.7676606 none
334 Administrative and specialized secretaries	2.410127 333.7196510 none
351 ICT operations and user support technicians	2.474549 305.6052807 none
241 Accountants, financial analysts and fund managers	2.534461 251.4058600 none
335 Tax and related government associate professionals	2.586686 250.8147705 none
515 Building caretakers and related workers	2.522386 250.0742967 none
321 Medical and pharmaceutical technicians	2.493038 230.0064060 none
213 Biologists, pharmacologists and specialists in agriculture and forestry	2.303600 228.0668837 none
134 Architectural and engineering managers	3.161068 226.5669850 none
333 Business services agents	3.001597 222.9774594 none
411 Office assistants and other secretaries	2.227235 214.6750980 none
243 Marketing and public relations professionals	1.481519 179.3099186 none
264 Authors, journalists and linguists	2.046538 164.8394018 none
129 Administration and service managers not elsewhere classified	4.059900 158.7687951 none
342 Athletes, fitness instructors and recreational workers	1.586943 132.1816646 none
159 Other social services managers	2.541205 69.9268014 none
123 Administration and planning managers	3.849200 50.0164438 none
541 Other surveillance and security workers	2.460130 41.9287342 none
235 Teaching professionals not elsewhere classified	1.415591 40.1919620 none
911 Cleaners and helpers	1.938366 35.3850213 none
533 Health care assistants	2.157379 22.0488822 none
534 Attendants, personal assistants and related workers	1.959595 21.6985964 none
332 Insurance advisers, sales and purchasing agents	2.637486 19.1041252 none
131 Information and communications technology service managers	4.000609 17.0841502 none
311 Physical and engineering science technicians	2.325958 16.5471030 none
214 Engineering professionals	2.626260 15.6152029 none

ssyk	Increase in salary	F-value interaction
432 Stores and transport clerks	1.231854	13.9155189 none
723 Machinery mechanics and fitters	2.362984	12.5847785 none
532 Personal care workers in health services	2.906578	6.3336230 none
512 Cooks and cold-buffet managers	2.483280	5.4612144 none
732 Printing trades workers	2.158854	5.2405535 none
234 Primary- and pre-school teachers	2.985653	4.6801860 none
422 Client information clerks	2.527801	1.7881181 none
531 Child care workers and teachers aides	1.881615	0.8334164 none
611 Market gardeners and crop growers	1.980288	0.3732826 none
341 Social work and religious associate professionals	2.357787	0.0276157 none
941 Fast-food workers, food preparation assistants	1.981512	0.0046670 none
<pre>merge(summary_table, anova_table, c("ssyk", "interact filter (term.x == "year_n") %>% filter (contcol.y > 0) %>% filter (interaction == "sector and sex") %>% mutate (estimate = (exp(estimate) - 1) * 100) %>% select (ssyk, estimate, statistic.y, interaction) % rename (`F-value` = statistic.y) %>% rename (`Increase in salary` = estimate) %>% arrange (desc (`F-value`)) %>% knitr::kable(booktabs = TRUE, caption = 'Correlation for F-value (sector and sector salaries')</pre>	}>%	

Table 3: Correlation for F-value (sector and sex) and the yearly increase in salaries

ssyk	Increase in salary	F-value interaction
911 Cleaners and helpers	1.966955	183.3539258 sector and sex
331 Financial and accounting associate professionals	1.988628	84.3723061 sector and sex
342 Athletes, fitness instructors and recreational workers	1.609407	53.2856268 sector and sex
351 ICT operations and user support technicians	2.474549	53.1541368 sector and sex
241 Accountants, financial analysts and fund managers	2.549161	39.5233707 sector and sex
333 Business services agents	2.952973	32.5926396 sector and sex
611 Market gardeners and crop growers	1.936338	25.7248443 sector and sex
243 Marketing and public relations professionals	1.489860	24.3260135 sector and sex
533 Health care assistants	2.157379	19.0714901 sector and sex
732 Printing trades workers	2.207617	16.8065594 sector and sex
512 Cooks and cold-buffet managers	2.497408	15.7408059 sector and sex

ssyk	Increase in salary	F-value interaction
159 Other social services managers	2.511558	15.1529257 sector and sex
123 Administration and planning managers	3.819995	14.2436097 sector and sex
532 Personal care workers in health services	2.899482	13.1974037 sector and sex
334 Administrative and specialized secretaries	2.416957	12.8317750 sector and sex
213 Biologists, pharmacologists and specialists in agriculture and forestry	2.317893	10.2339741 sector and sex
134 Architectural and engineering managers	3.161068	9.7816667 sector and sex
321 Medical and pharmaceutical technicians	2.493038	8.6670096 sex
941 Fast-food workers, food preparation assistants	1.989465	8.6486138 sector and sex
335 Tax and related government associate professionals	2.586686	8.4427211 sector and sex
242 Organisation analysts, policy administrators and human resource specialists	1.594999	8.1271264 sector and sex
723 Machinery mechanics and fitters	2.337983	6.4299201 sector and sex
332 Insurance advisers, sales and purchasing agents	2.627736	5.7400150 sector and sex
311 Physical and engineering science technicians	2.316138	5.3473607 sector and sex
234 Primary- and pre-school teachers	2.999161	5.0012138 sector and sex
129 Administration and service managers not elsewhere classified	4.121066	4.6687219 sector and sex
819 Process control technicians	2.250895	3.7631281 sector and sex
534 Attendants, personal assistants and related workers	1.959595	3.5292601 sector and sex
131 Information and communications technology service managers	4.029303	2.5233240 sector and sex
264 Authors, journalists and linguists	2.033203	2.5032667 sector and sex
341 Social work and religious associate professionals	2.357787	2.2988889 sector and sex
251 ICT architects, systems analysts and test managers	2.217446	2.1823519 sector and sex
422 Client information clerks	2.519631	2.1300830 sector and sex
432 Stores and transport clerks	1.231854	1.5040767 sector and sex
541 Other surveillance and security workers	2.457007	1.3945913 sex
214 Engineering professionals	2.629057	1.3225970 sector and sex

ssyk	Increase in salary	F-value	interaction
962 Newspaper distributors, janitors and other service workers	1.971413	0.3480210	sector and sex
531 Child care workers and teachers aides	1.879581	0.1432868	sector and sex
515 Building caretakers and related workers	2.522844	0.1379141	sector and sex
235 Teaching professionals not elsewhere classified	1.424488	0.0670314	sector and sex
411 Office assistants and other secretaries	2.227235	0.0001653	sector and sex
<pre>merge(summary_table, anova_table, c("ssyk", "intera filter (term.x == "year_n") %>% filter (contcol.y > 0) %>% filter (interaction == "sector and edulevel") %>% mutate (estimate = (exp(estimate) - 1) * 100) %>% select (ssyk, estimate, statistic.y, interaction) rename (`F-value` = statistic.y) %>% rename (`Increase in salary` = estimate) %>% arrange (desc (`F-value`)) %>% knitr::kable(booktabs = TRUE, caption = 'Correlation for F-value (sector and increase in salaries')</pre>	%>%		
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Table 4: Correlation for F-value (sector and edulevel) and the yearly increase in salaries

ssyk	Increase in F-value interaction salary
335 Tax and related government associate professionals	2.586686 48.2902588 sector and edulevel
234 Primary- and pre-school teachers	2.936162 32.3560802 sector and edulevel
214 Engineering professionals	2.636843 29.4462376 sector and edulevel
332 Insurance advisers, sales and purchasing agents	2.420632 26.7105578 sector and edulevel
432 Stores and transport clerks	1.231854 24.9194979 sector and edulevel
134 Architectural and engineering managers	3.161068 20.3665858 sector and edulevel
321 Medical and pharmaceutical technicians	2.493038 18.1556092 sector and edulevel
723 Machinery mechanics and fitters	2.358171 17.3272819 sector and edulevel
311 Physical and engineering science technicians	2.330300 12.9686992 sector and edulevel
123 Administration and planning managers	3.809019 11.7436022 sector and edulevel
732 Printing trades workers	2.220728 11.0453452 sector and edulevel
241 Accountants, financial analysts and fund managers	2.647068 10.5100504 sector and edulevel

	Increase in	
ssyk	salary	F-value interaction
235 Teaching professionals not elsewhere classified	1.396076	8.6264575 sector and edulevel
213 Biologists, pharmacologists and specialists in agriculture and forestry	2.264110	7.7952190 sector and edulevel
941 Fast-food workers, food preparation assistants	1.981512	7.5766373 sector and edulevel
331 Financial and accounting associate professionals	1.980338	6.9758948 sector and edulevel
532 Personal care workers in health services	2.919933	6.6231457 sector and edulevel
534 Attendants, personal assistants and related workers	1.959595	5.7464596 sector and edulevel
962 Newspaper distributors, janitors and other service workers	1.971413	5.4435379 sector and edulevel
131 Information and communications technology service managers	3.908111	4.9661376 sector and edulevel
159 Other social services managers	2.565523	4.7197836 sector and edulevel
129 Administration and service managers not elsewhere classified	4.160995	4.3062655 sector and edulevel
251 ICT architects, systems analysts and test managers	2.209131	3.5539841 sector and edulevel
512 Cooks and cold-buffet managers	2.435903	3.4242806 sector and edulevel
333 Business services agents	2.955970	3.3635706 sector and edulevel
243 Marketing and public relations professionals	1.473064	3.2639790 sector and edulevel
264 Authors, journalists and linguists	2.043677	3.0449469 sector and edulevel
334 Administrative and specialized secretaries	2.387467	2.8528940 sector and edulevel
422 Client information clerks	2.527801	2.3007696 sector and edulevel
242 Organisation analysts, policy administrators and human resource specialists	1.612900	1.9599904 sector and edulevel
351 ICT operations and user support technicians	2.474549	1.5903279 sector and edulevel
819 Process control technicians	2.250895	1.5586112 sector and edulevel
341 Social work and religious associate professionals	2.357787	1.3263497 sector and edulevel
611 Market gardeners and crop growers	2.016345	1.2719035 sector and edulevel
541 Other surveillance and security workers	2.460130	1.0672924 sector and edulevel
411 Office assistants and other secretaries	2.227235	1.0114046 sector and edulevel
515 Building caretakers and related workers	2.526100	0.8254389 sector and edulevel

ssyk	Increase in salary	F-value interaction
342 Athletes, fitness instructors and recreational workers	1.540952	0.8230839 sector and edulevel
531 Child care workers and teachers aides	1.897863	0.7969406 sector and edulevel
533 Health care assistants	2.157379	0.5425248 sector and edulevel
911 Cleaners and helpers	1.938366	0.0965584 sector and edulevel
<pre>merge(summary_table, anova_table, c("ssyk", "interfilter (term.x == "year_n") %>% filter (contcol.y > 0) %>% filter (interaction == "sector and year") %>% mutate (estimate = (exp(estimate) - 1) * 100) % select (ssyk, estimate, statistic.y, interaction rename (`F-value` = statistic.y) %>% rename (`Increase in salary` = estimate) %>% arrange (desc (`F-value`)) %>% knitr::kable(booktabs = TRUE, caption = 'Correlation for F-value (sector and in salaries')</pre>	>% n) %>%	

Table 5: Correlation for F-value (sector and year) and the yearly increase in salaries

ssyk	Increase in salary	F-value interaction
129 Administration and service managers not elsewhere classified	5.8528187 1	7.1667457 sector and year
351 ICT operations and user support technicians	3.2455362 1	6.6101284 sector and year
334 Administrative and specialized secretaries	1.0672775 1	4.9508269 sector and year
534 Attendants, personal assistants and related workers	2.2769477 1	sector and 2.4036331 year
422 Client information clerks	3.2733483 1	1.6812523 sector and year
962 Newspaper distributors, janitors and other service workers	2.4907346 1	1.5673142 sector and year
264 Authors, journalists and linguists	2.9905121 1	0.5945994 sector and year
531 Child care workers and teachers aides	2.4510286	9.5936100 sector and year
242 Organisation analysts, policy administrators and human resource specialists	2.3355826	7.7816291 sector and year
432 Stores and transport clerks	0.5332567	7.4196762 sector and year
243 Marketing and public relations professionals	2.1828057	5.5660180 sector and year
732 Printing trades workers	2.8825380	4.3716553 sector and year
213 Biologists, pharmacologists and specialists in agriculture and forestry	2.8099783	3.6786224 sector and year

ssyk	Increase in salary	F-value interaction
611 Market gardeners and crop growers	•	3.3017939 sector and year
131 Information and communications technology service managers	3.3504930	2.8750373 sector and year
532 Personal care workers in health services	2.7411030	2.8290655 sector and year
235 Teaching professionals not elsewhere classified	1.8047571	2.7841902 sector and year
311 Physical and engineering science technicians	2.8608741	2.6173373 sector and year
533 Health care assistants	1.9920113	2.2486864 sector and year
214 Engineering professionals	3.0209963	2.0446973 sector and year
723 Machinery mechanics and fitters	2.6258864	1.4272778 sector and year
515 Building caretakers and related workers	2.3423357	1.3492650 sector and year
321 Medical and pharmaceutical technicians	2.1376334	1.2494762 sector and year
941 Fast-food workers, food preparation assistants	2.1906312	sector and 1.1458297 year
234 Primary- and pre-school teachers	3.1522849	0.8075167 sector and year
411 Office assistants and other secretaries	2.4217447	0.7799156 sector and year
241 Accountants, financial analysts and fund managers	2.2625663	0.7751592 sector and year
541 Other surveillance and security workers	2.3018747	0.6646888 sector and year
134 Architectural and engineering managers	3.3811821	0.5852232 sector and year
333 Business services agents	3.2520012	0.5424390 sector and year
335 Tax and related government associate professionals	2.3986233	0.3139375 sector and year
123 Administration and planning managers	4.0945884	0.2682143 sector and year
911 Cleaners and helpers	2.0082873	0.2021948 sector and year
341 Social work and religious associate professionals	2.3994292	0.0754320 sector and year
342 Athletes, fitness instructors and recreational workers	1.6673896	0.0633281 sector and year
512 Cooks and cold-buffet managers	2.4249631	0.0494622 sector and year
819 Process control technicians	2.2225873	0.0378303 sector and year
331 Financial and accounting associate professionals	2.0557505	0.0377336 sector and year

ssyk	Increase in salary	F-value interaction
251 ICT architects, systems analysts and test managers	2.2536138	0.0294035 sector and year
332 Insurance advisers, sales and purchasing agents	2.6015664	0.0125275 sector and year
159 Other social services managers	2.5487758	0.0013436 sector and year
<pre>merge(summary_table, anova_table, c("ssyk", "intera- filter (term.x == "year_n") %>% filter (contcol.y > 1) %>% filter (interaction == "sector, year, edulevel and filter (!(contcol.y < 3 & interaction == "sector," %>%</pre>	d sex") %>	8
<pre>mutate (estimate = (exp(estimate) - 1) * 100) %>% select (ssyk, estimate, statistic.y, interaction) rename (`F-value` = statistic.y) %>% rename (`Increase in salary` = estimate) %>% arrange (desc (`F-value`)) %>% knitr::kable(booktabs = TRUE, caption = 'Correlation for F-value (sector, yearly increase in salaries')</pre>	%>%	l and sex) and the

Increase in

Table 6: Correlation for F-value (sector, year, edulevel and sex) and the yearly increase in salaries

ssyk	Increase in salary	F-value	interaction
264 Authors, journalists and linguists	2.0717871	5.0646298	sector, year, edulevel and sex
311 Physical and engineering science technicians	-1.9358596	3.3593399	sector, year, edulevel and sex
159 Other social services managers	2.1339049	2.5206357	sector, year, edulevel and sex
134 Architectural and engineering managers	-0.2221716	2.4150753	sector, year, edulevel and sex
331 Financial and accounting associate professionals	0.2989757	2.2892768	sector, year, edulevel and sex
342 Athletes, fitness instructors and recreational workers	2.7799167	2.1976399	sector, year, edulevel and sex
214 Engineering professionals	5.2732904	1.9988132	sector, year, edulevel and sex
723 Machinery mechanics and fitters	2.1334983	1.8843643	sector, year, edulevel and sex
432 Stores and transport clerks	-0.2883737	1.8215339	sector, year, edulevel and sex
241 Accountants, financial analysts and fund managers	2.6268377	1.8184489	sector, year, edulevel and sex
533 Health care assistants	1.3176280	1.6357939	sector, year, edulevel and sex
911 Cleaners and helpers	1.1875630	1.6319874	sector, year, edulevel and sex
129 Administration and service managers not elsewhere classified	16.0932403	1.5716145	sector, year, edulevel and sex

ssyk	Increase in salary	F-value	interaction
512 Cooks and cold-buffet managers	2.2335677	1.4882138	sector, year, edulevel and sex
242 Organisation analysts, policy administrators and human resource specialists	1.9377802	1.4494844	sector, year, edulevel and sex
234 Primary- and pre-school teachers	2.6286211	1.4408203	sector, year, edulevel and sex
235 Teaching professionals not elsewhere classified	1.4806649	1.4162230	sector, year, edulevel and sex
532 Personal care workers in health services	3.0809717	1.3439893	sector, year, edulevel and sex
962 Newspaper distributors, janitors and other service workers	1.7217534 ⁻	1.3344044	sector, year, edulevel and sex
332 Insurance advisers, sales and purchasing agents	4.5515134 (0.9604905	sector, year, edulevel and sex
123 Administration and planning managers	1.6860436 (0.9155189	sector, year, edulevel and sex
541 Other surveillance and security workers	3.1372549	0.9113280	sector, year, edulevel and sex
213 Biologists, pharmacologists and specialists in agriculture and forestry	2.0956711 (0.9044780	sector, year, edulevel and sex
422 Client information clerks	7.6231100	0.8611877	sector, year, edulevel and sex
515 Building caretakers and related workers	3.8284291 (0.8343329	sector, year, edulevel and sex
334 Administrative and specialized secretaries	-1.8021942 (0.7940149	sector, year, edulevel and sex
531 Child care workers and teachers aides	2.9556278 (0.7580341	sector, year, edulevel and sex
131 Information and communications technology service managers	6.2536436 (0.7429188	sector, year, edulevel and sex
243 Marketing and public relations professionals	-0.7939038 (0.7338409	sector, year, edulevel and sex
351 ICT operations and user support technicians	3.2312302 (0.7188959	sector, year, edulevel and sex
732 Printing trades workers	3.9993823 (0.6905319	sector, year, edulevel and sex
321 Medical and pharmaceutical technicians	2.2164674 (0.6326293	sector, year, edulevel and sex
611 Market gardeners and crop growers	2.2753348 (0.6157880	sector, year, edulevel and sex
341 Social work and religious associate professionals	3.0666620 (0.5266982	sector, year, edulevel and sex
941 Fast-food workers, food preparation assistants	2.7940568 (0.4253642	sector, year, edulevel and sex
335 Tax and related government associate professionals	5.0470456	0.3217115	sector, year, edulevel and sex
251 ICT architects, systems analysts and test managers	2.9031256 (0.2684425	sector, year, edulevel
819 Process control technicians	1.6741717 (0.2603997	sector, year, edulevel and sex

ssyk	F-value interaction	
333 Business services agents	1.8143344 0.1540308 sector, year, edulevel and sex	
411 Office assistants and other secretaries	5.5231019 0.0911334 sector, year, edulevel and sex	
534 Attendants, personal assistants and related workers	2.6927719 0.0899580 sector, year, edulevel	

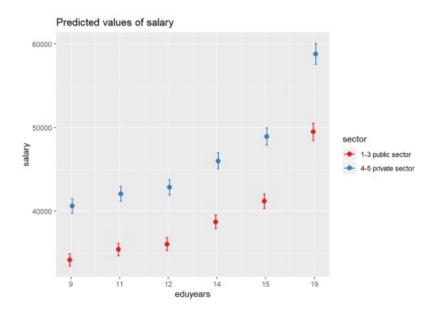
Let's check what we have found.

```
temp <- tbnum %>%
  filter(`occuptional (SSYK 2012)` == "242 Organisation analysts, policy
administrators and human resource specialists")

model <- lm (log(salary) ~ year_n + eduyears + sector + sex, data = temp)

plot_model(model, type = "pred", terms = c("eduyears", "sector"))

## Model has log-transformed response. Back-transforming predictions to original</pre>
```



response scale. Standard errors are still on the log-scale.

Figure 3: Highest F-value sector, Organisation analysts, policy administrators and human resource specialists

```
temp <- tbnum %>%
  filter(`occuptional (SSYK 2012)` == "941 Fast-food workers, food preparation
assistants")

model <-lm (log(salary) ~ year_n + eduyears + sector + sex, data = temp)

plot_model(model, type = "pred", terms = c("eduyears", "sector"))

## Model has log-transformed response. Back-transforming predictions to original response scale. Standard errors are still on the log-scale.</pre>
```

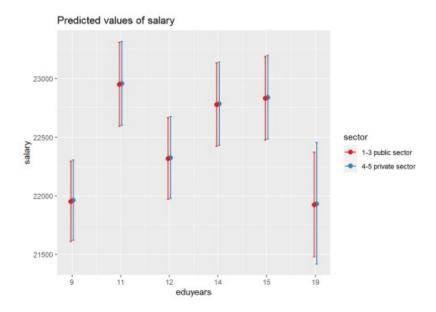


Figure 4: Lowest F-value sector, Fast-food workers, food preparation assistants

```
temp <- tbnum %>%
  filter(`occuptional (SSYK 2012)` == "911 Cleaners and helpers")

model <- lm (log(salary) ~ year_n + eduyears + sector * sex, data = temp)

plot_model(model, type = "pred", terms = c("eduyears", "sex", "sector"))

## Model has log-transformed response. Back-transforming predictions to original response scale. Standard errors are still on the log-scale.</pre>
```

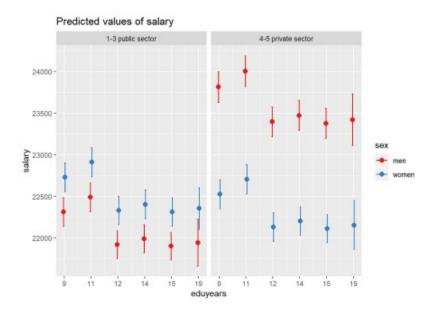


Figure 5: Highest F-value interaction sector and gender, Cleaners and helpers

```
temp <- tbnum %>%
  filter(`occuptional (SSYK 2012)` == "411 Office assistants and other
secretaries")

model <- lm (log(salary) ~ year_n + eduyears + sector * sex, data = temp)

plot_model(model, type = "pred", terms = c("eduyears", "sex", "sector"))

## Model has log-transformed response. Back-transforming predictions to original</pre>
```

response scale. Standard errors are still on the log-scale.

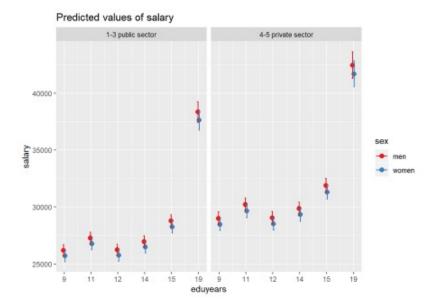


Figure 6: Lowest F-value interaction sector and gender, Office assistants and other secretaries

```
temp <- tbnum %>%
  filter(`occuptional (SSYK 2012)` == "335 Tax and related government associate
professionals")

model <- lm (log(salary) ~ year_n + eduyears * sector + sex, data = temp)

plot_model(model, type = "pred", terms = c("eduyears", "sector"))

## Warning in predict.lm(model, newdata = fitfram, type = "response", se.fit =
## se, : prediction from a rank-deficient fit may be misleading

## Model has log-transformed response. Back-transforming predictions to original
response scale. Standard errors are still on the log-scale.</pre>
```

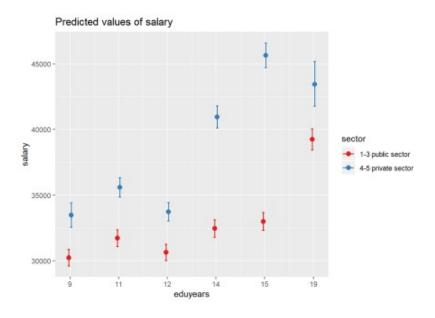


Figure 7: Highest F-value interaction sector and edulevel, Tax and related government associate professionals

```
temp <- tbnum %>%
  filter(`occuptional (SSYK 2012)` == "911 Cleaners and helpers")
```

```
model <- lm (log(salary) ~ year_n + eduyears * sector + sex, data = temp)
plot_model(model, type = "pred", terms = c("eduyears", "sector"))
## Warning in predict.lm(model, newdata = fitfram, type = "response", se.fit = ## se, : prediction from a rank-deficient fit may be misleading</pre>
```

Model has log-transformed response. Back-transforming predictions to original response scale. Standard errors are still on the log-scale.

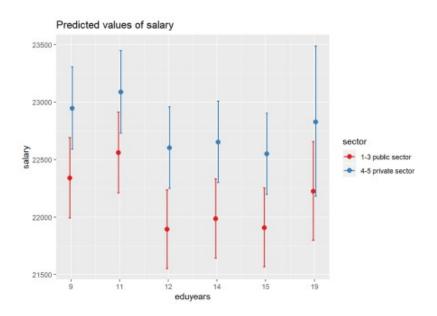


Figure 8: Lowest F-value interaction sector and edulevel, Cleaners and helpers

```
temp <- tbnum %>%
  filter(`occuptional (SSYK 2012)` == "129 Administration and service managers
not elsewhere classified")

model <- lm (log(salary) ~ year_n * sector + eduyears + sex, data = temp)

plot_model(model, type = "pred", terms = c("eduyears", "year_n", "sector"))

## Model has log-transformed response. Back-transforming predictions to original response scale. Standard errors are still on the log-scale.</pre>
```

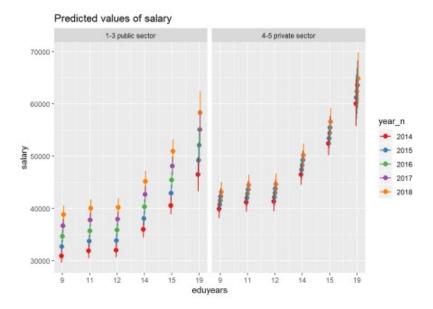


Figure 9: Highest F-value interaction sector and year, Administration and service managers not elsewhere classified

```
temp <- tbnum %>%
  filter(`occuptional (SSYK 2012)` == "159 Other social services managers")

model <- lm (log(salary) ~ year_n * sector + eduyears + sex, data = temp)

plot_model(model, type = "pred", terms = c("eduyears", "year_n", "sector"))

## Model has log-transformed response. Back-transforming predictions to original response scale. Standard errors are still on the log-scale.</pre>
```

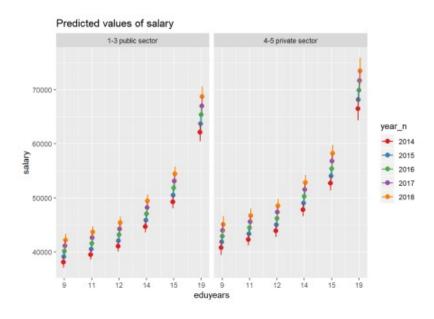


Figure 10: Lowest F-value interaction sector and year, Other social services managers

```
temp <- tbnum %>%
  filter(`occuptional (SSYK 2012)` == "264 Authors, journalists and linguists")
model <- lm (log(salary) ~ year_n * eduyears * sector * sex, data = temp)
plot_model(model, type = "pred", terms = c("eduyears", "year_n", "sex",
"sector"))</pre>
```

Warning in predict.lm(model, newdata = fitfram, type = "response", se.fit =
se, : prediction from a rank-deficient fit may be misleading

Model has log-transformed response. Back-transforming predictions to original response scale. Standard errors are still on the log-scale.

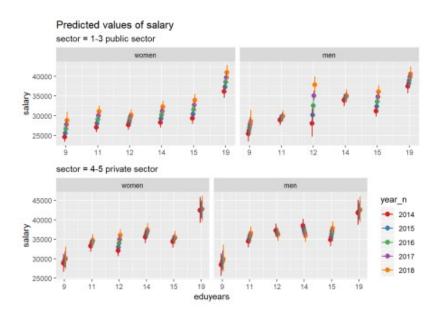


Figure 11: Highest F-value interaction sector, edulevel, year and gender, Authors, journalists and linguists

```
## TableGrob (2 x 1) "arrange": 2 grobs
## z cells name grob
## 1 1 (1-1,1-1) arrange gtable[layout]
## 2 2 (2-2,1-1) arrange gtable[layout]

temp <- tbnum %>%
    filter(`occuptional (SSYK 2012)` == "534 Attendants, personal assistants and related workers")

model <- lm (log(salary) ~ year_n * eduyears * sector * sex, data = temp)

plot_model(model, type = "pred", terms = c("eduyears", "year_n", "sex", "sector"))

## Warning in predict.lm(model, newdata = fitfram, type = "response", se.fit = ## se, : prediction from a rank-deficient fit may be misleading

## Model has log-transformed response. Back-transforming predictions to original</pre>
```

response scale. Standard errors are still on the log-scale.

Predicted values of salary

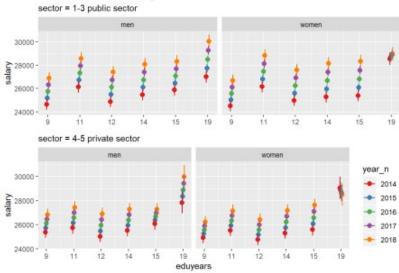


Figure 12: Lowest F-value interaction sector, edulevel, year and gender, Attendants, personal assistants and related workers

```
## TableGrob (2 x 1) "arrange": 2 grobs
## z cells name grob
## 1 1 (1-1,1-1) arrange gtable[layout]
## 2 2 (2-2,1-1) arrange gtable[layout]
```