

IGP DATA ANALYSIS

PROJECT TEAM 1

August 22

Data Description: This dataset is part of a larger dataset that has been collected to help to know the impact of Covid 19 Pandemic on the employment profession in the 4 regions of the UK and to also observe the various shifts that occurred in different professions as a result of this, paying particular attention to professions that gained, loss and constant.

```
source("mypackages.R")
```

Loading required package: kableExtra

Attaching package: 'flextable'

The following objects are masked from 'package:kableExtra':

as_image, footnote

Attaching package: 'dplyr'

The following object is masked from 'package:kableExtra':

group_rows

The following objects are masked from 'package:stats':

filter, lag

The following objects are masked from 'package:base':

intersect, setdiff, setequal, union

Attaching package: 'lubridate'

The following objects are masked from 'package:base':

date, intersect, setdiff, union

Attaching package: 'MASS'

The following object is masked from 'package:dplyr':

select

Attaching package: 'psych'

The following objects are masked from 'package:ggplot2':

%%, alpha

Attaching package: 'patchwork'

The following object is masked from 'package:MASS':

area

corrplot 0.92 loaded

```
source("helperFunctions.R")
```

```
library(tidyr)
```

```
# load dataset
```

```
employment_data <- read.csv("merged_data.csv")
```

```
#glimpse the loaded csv file to check structure of dataset such as no of rows, no of column and variable
```

```
library(dplyr)
```

Data Preparation

We are interested in modelling the employment by professions where Country is equal to Scotland.

Rows of data with this property are selected and used to perform the rest of the analysis.

```
#Select rows that satisfies the condition specified in the modelling
```

```
employ_scotland<-subset(employment_data,employment_data$Country == "Scotland")
```

```
tinytex::install_tinytex()
```

Found 'C:\Users\Haywh\AppData\Roaming\TinyTeX\bin\win32\tlmgr.bat', which indicates a LaTeX distribution

Error in tinytex::install_tinytex(): If you want to force installing TinyTeX anyway, use tinytex::install_tinytex(TRUE)

```
glimpse(employ_scotland)
```

Rows: 18

Columns: 17

| | |
|--|---------------|
| \$ Date | <int> 2004, ~ |
| \$ Country | <chr> "Scotl~ |
| \$ X11.Corporate.managers.and.directors | <int> 132700~ |
| \$ X21.Science..research..engineering.and.technology.professionals | <int> 100400~ |
| \$ X22.Health.professionals | <int> 94100,~ |
| \$ X23.Teaching.and.educational.professionals | <int> 108300~ |
| \$ X24.Business..media.and.public.service.professionals | <int> 101200~ |

```

$ X33.Protective.service.occupations <int> 33500,~
$ X34.Culture..media.and.sports.occupations <int> 28500,~
$ X41.Administrative.occupations <int> 239700~
$ X42.Secretarial.and.related.occupations <int> 73500,~
$ X51.Skilled.agricultural.and.related.trades <int> 34000,~
$ X53.Skilled.construction.and.building.trades <int> 93200,~
$ X61.Caring.personal.service.occupations <int> 159900~
$ X62.Leisure..travel.and.related.personal.service.occupations <int> 48100,~
$ X81.Process..plant.and.machine.operatives <int> 96900,~
$ X82.Transport.and.mobile.machine.drivers.and.operatives <int> 91600,~

```

The following exploratory data analysis will be done using descriptive statistics, exploratory graphs and linear correlation.

summary(): To measure central tendency by checking for variation in mean, median, also for outliers. If variation or difference between mean and median is much then it means there's outliers.

Skimr: this is another method used for summarizing data. It gives a broad set of summary statistics with a single function call.

```
summary(employ_scotland)
```

```

      Date      Country      X11.Corporate.managers.and.directors
Min.   :2004   Length:18   Min.     :132700
1st Qu.:2008   Class :character 1st Qu.:139375
Median :2012   Mode  :character Median :148400
Mean   :2012                                     Mean   :147511
3rd Qu.:2017                                     3rd Qu.:150050
Max.   :2021                                     Max.   :169300
X21.Science..research..engineering.and.technology.professionals
Min.     :100400
1st Qu.:112300
Median :125200
Mean    :132767
3rd Qu.:151425
Max.    :197700
X22.Health.professionals X23.Teaching.and.educational.professionals
Min.    : 94100          Min.     :108300
1st Qu.:112425          1st Qu.:115325
Median :120350          Median :116650
Mean    :119828          Mean    :119744
3rd Qu.:130150          3rd Qu.:125350
Max.    :148000          Max.    :133200
X24.Business..media.and.public.service.professionals
Min.     : 98900
1st Qu.:110150
Median :121450
Mean    :123606
3rd Qu.:132500
Max.    :163800
X33.Protective.service.occupations X34.Culture..media.and.sports.occupations
Min.     :26900          Min.     :28500
1st Qu.:33750          1st Qu.:38200
Median :35950          Median :42300

```

| | |
|--|---------------|
| Mean :35289 | Mean :44144 |
| 3rd Qu.:36750 | 3rd Qu.:50475 |
| Max. :40900 | Max. :64900 |
| X41.Administrative.occupations X42.Secretarial.and.related.occupations | |
| Min. :200900 | Min. :39400 |
| 1st Qu.:215525 | 1st Qu.:50700 |
| Median :218100 | Median :55950 |
| Mean :221078 | Mean :57900 |
| 3rd Qu.:226625 | 3rd Qu.:69650 |
| Max. :239700 | Max. :74000 |
| X51.Skilled.agricultural.and.related.trades | |
| Min. :29400 | |
| 1st Qu.:33700 | |
| Median :35800 | |
| Mean :34972 | |
| 3rd Qu.:36200 | |
| Max. :39900 | |
| X53.Skilled.construction.and.building.trades | |
| Min. : 59900 | |
| 1st Qu.: 87200 | |
| Median : 90200 | |
| Mean : 89778 | |
| 3rd Qu.: 94550 | |
| Max. :108600 | |
| X61.Caring.personal.service.occupations | |
| Min. :159900 | |
| 1st Qu.:179525 | |
| Median :186650 | |
| Mean :185739 | |
| 3rd Qu.:194200 | |
| Max. :203700 | |
| X62.Leisure..travel.and.related.personal.service.occupations | |
| Min. :48100 | |
| 1st Qu.:51425 | |
| Median :52950 | |
| Mean :54350 | |
| 3rd Qu.:57750 | |
| Max. :61600 | |
| X81.Process..plant.and.machine.operatives | |
| Min. :58600 | |
| 1st Qu.:64025 | |
| Median :69800 | |
| Mean :72633 | |
| 3rd Qu.:80925 | |
| Max. :96900 | |
| X82.Transport.and.mobile.machine.drivers.and.operatives | |
| Min. : 77300 | |
| 1st Qu.: 92925 | |
| Median : 99350 | |
| Mean : 97356 | |
| 3rd Qu.:101350 | |
| Max. :107200 | |

```
is.null(employ_scotland)
```

```
[1] FALSE
```

```
skimr::skim(employ_scotland)
```

```
Error in kable_latex(x = structure(c("Name", "Number of rows", "Number of columns", : unused argument (
```

Summary shows there are no outliers as variations between mean and median in all professions are minimal. We checked for null values and got a boolean FALSE.

EXPLORATORY GRAPHS

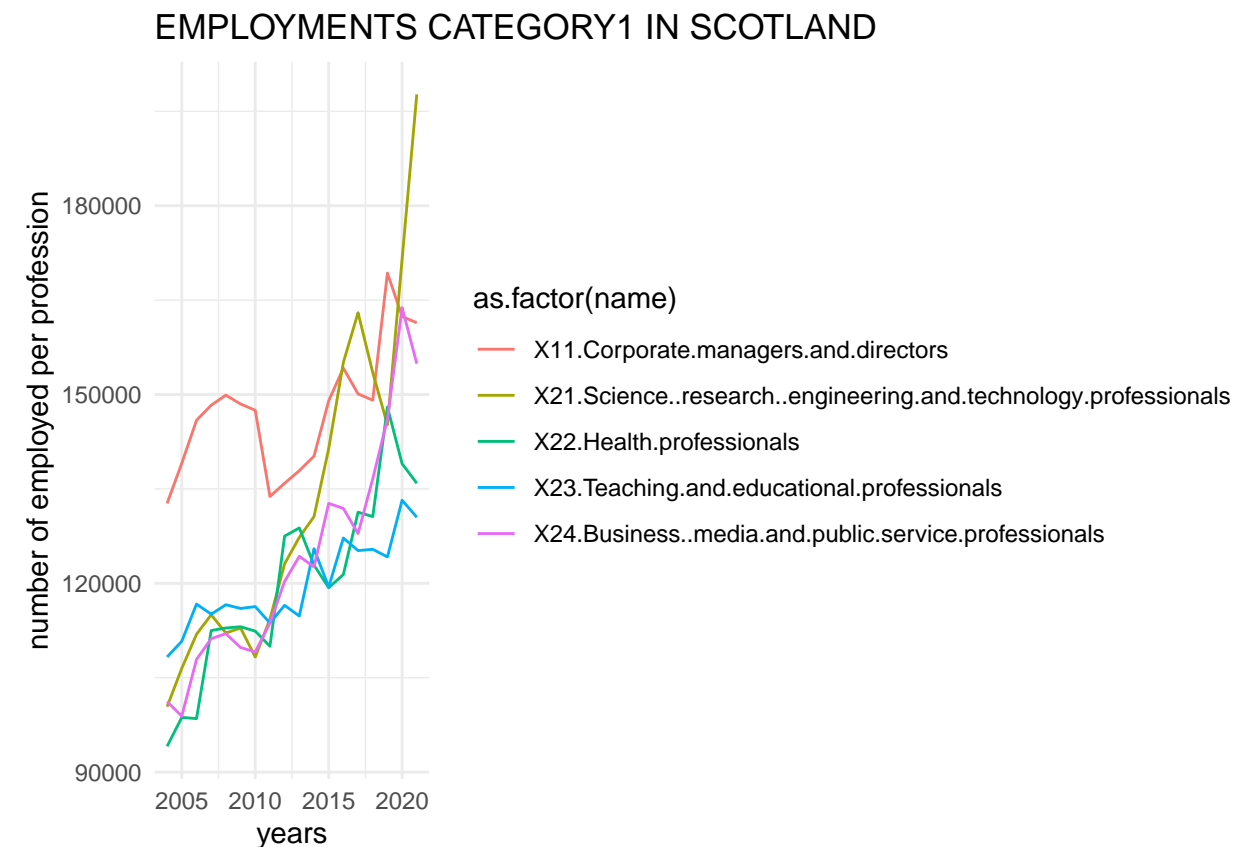
5 professions were selected and grouped into 3 categories and all visually analysed.

```
category1 <- employ_scotland[,c("Date", "Country", "X11.Corporate.managers.and.directors", "X21.Science..r
```

Here we make use of pivot longer which converts wide data into long data using the function pivot longer. Line graphs were plotted showing 5 professions in each categories.

```
#library(tidyverse)
category1_long <- category1%>%
  pivot_longer(
    cols = starts_with("X")
  )
```

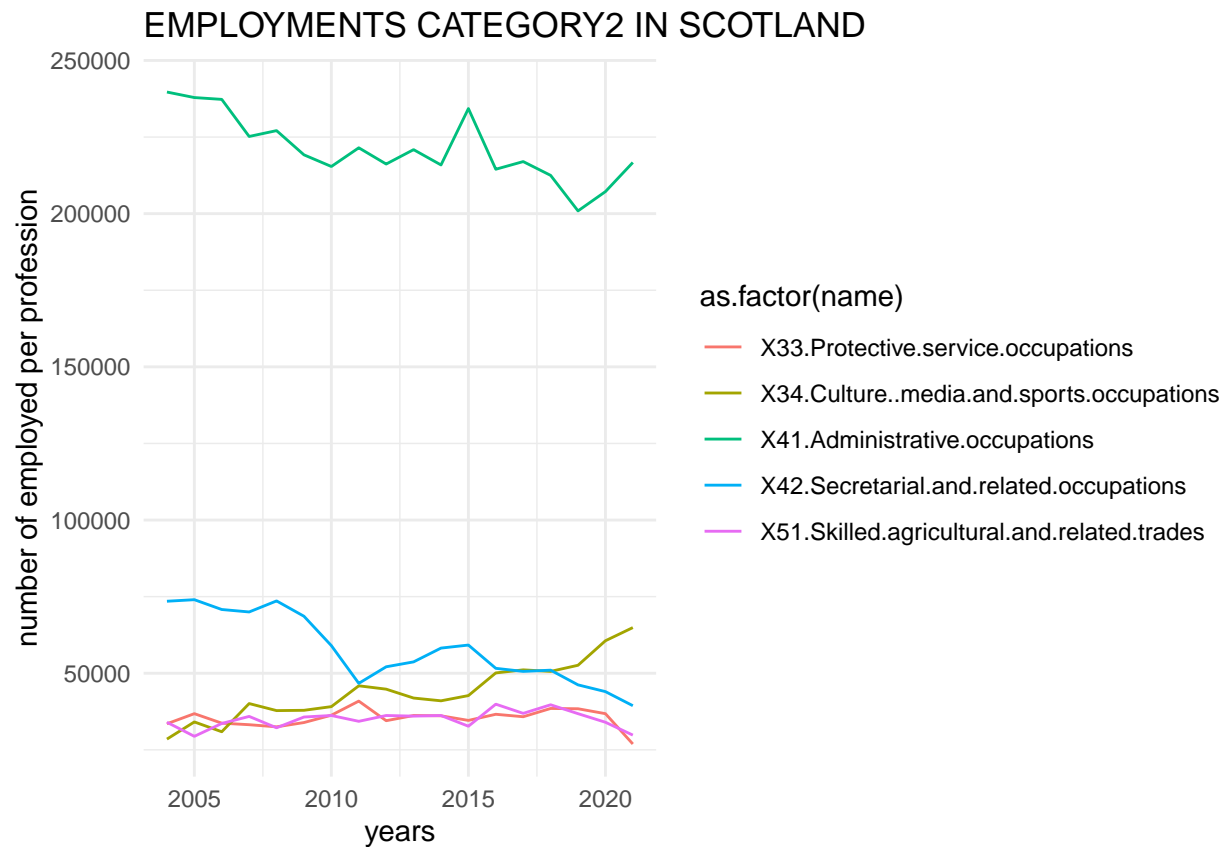
```
ggplot(category1_long, aes(x=Date, y=value, col=as.factor(name)))+geom_line()+xlab("years")+ylab("number of
```



```
category2 <- employ_scotland[,c("Date","Country","X33.Protective.service.occupations", "X34.Culture..me
```

```
category2_long <- category2%>%
  pivot_longer(
    cols = starts_with("X")
  )
```

```
ggplot(category2_long,aes(x=Date,y=value,col=as.factor(name)))+geom_line()+xlab("years")+ylab("number o
```

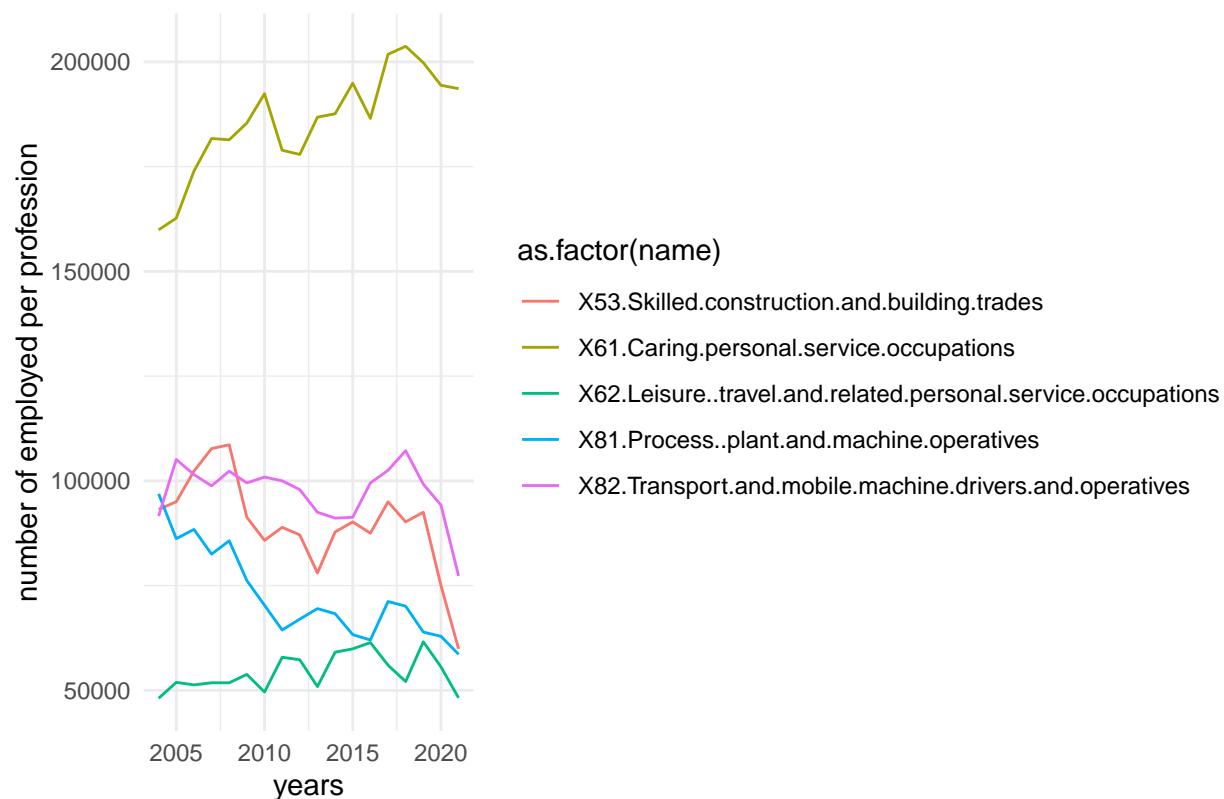


```
category3 <- employ_scotland[,c("Date","Country","X53.Skilled.construction.and.building.trades", "X61.C
```

```
category3_long <- category3%>%
  pivot_longer(
    cols = starts_with("X")
  )
```

```
ggplot(category3_long,aes(x=Date,y=value,col=as.factor(name)))+geom_line()+xlab("years")+ylab("number o
```

EMPLOYMENTS CATEGORY3 IN SCOTLAND



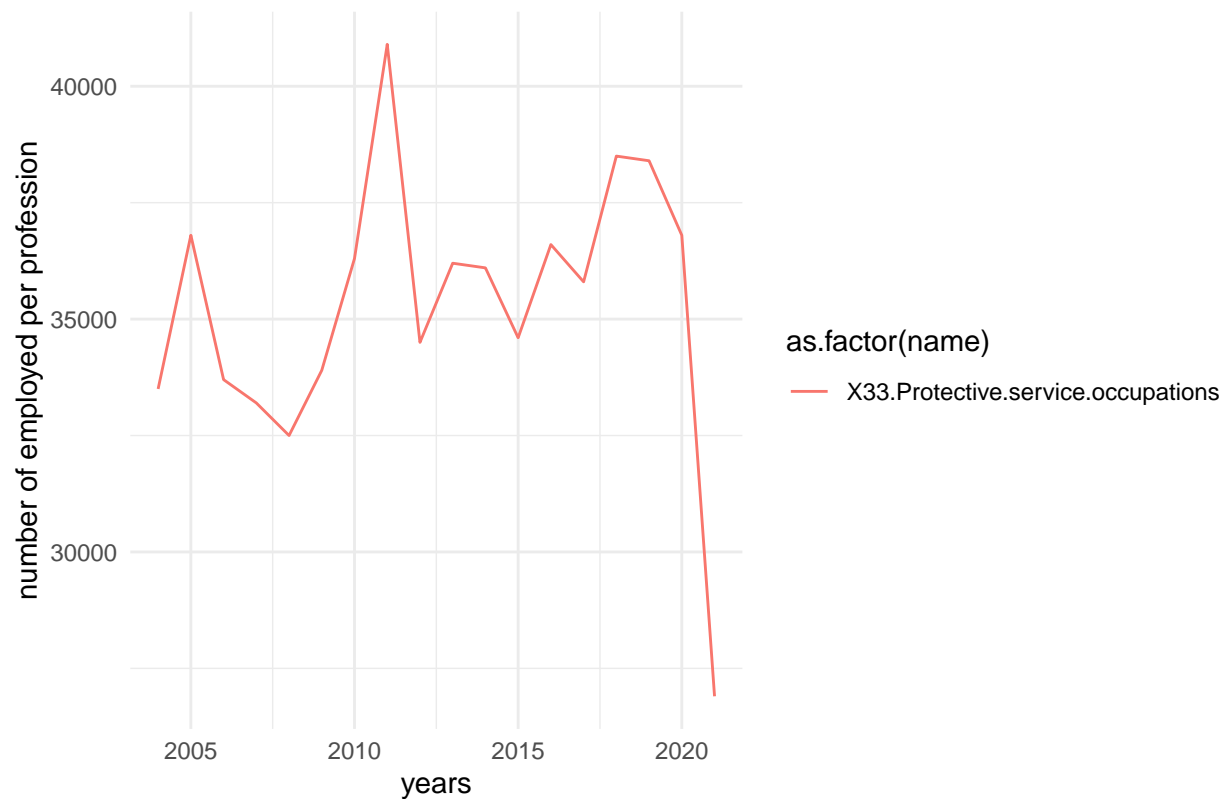
```
protective<- employ_scotland[,c("Date","Country","X33.Protective.service.occupations")]
```

```
protective2 <- protective%>%
  pivot_longer(
    cols = starts_with("X")
  )
```

From the categories plotted, professions with notable changes were plotted individually to further highlight the changes.

```
ggplot(protective2,aes(x=Date,y=value,col=as.factor(name)))+geom_line()+xlab("years")+ylab("number of e
```

PROTECTIVE PROFESSION IN SCOTLAND



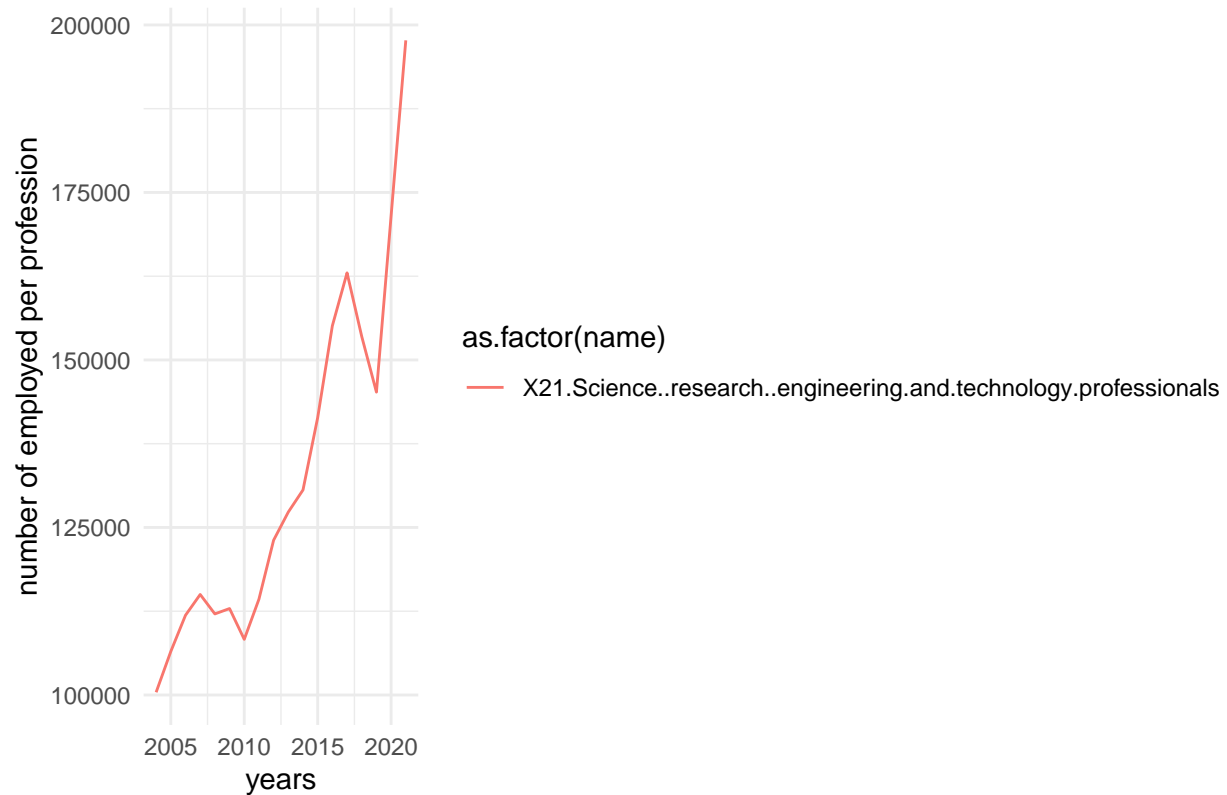
This shows there was a continual decline with effect initializing from 2019 but with significant COVID effect in 2020

```
science<- employ_scotland[,c("Date","Country","X21.Science..research..engineering.and.technology.profes
```

```
science2 <- science%>%
  pivot_longer(
    cols = starts_with("X")
  )
```

```
ggplot(science2,aes(x=Date,y=value,col=as.factor(name)))+geom_line()+xlab("years")+ylab("number of empl
```


SCIENCE, TECHNOLOGY AND ENGINEERING PROFESSION IN SCO

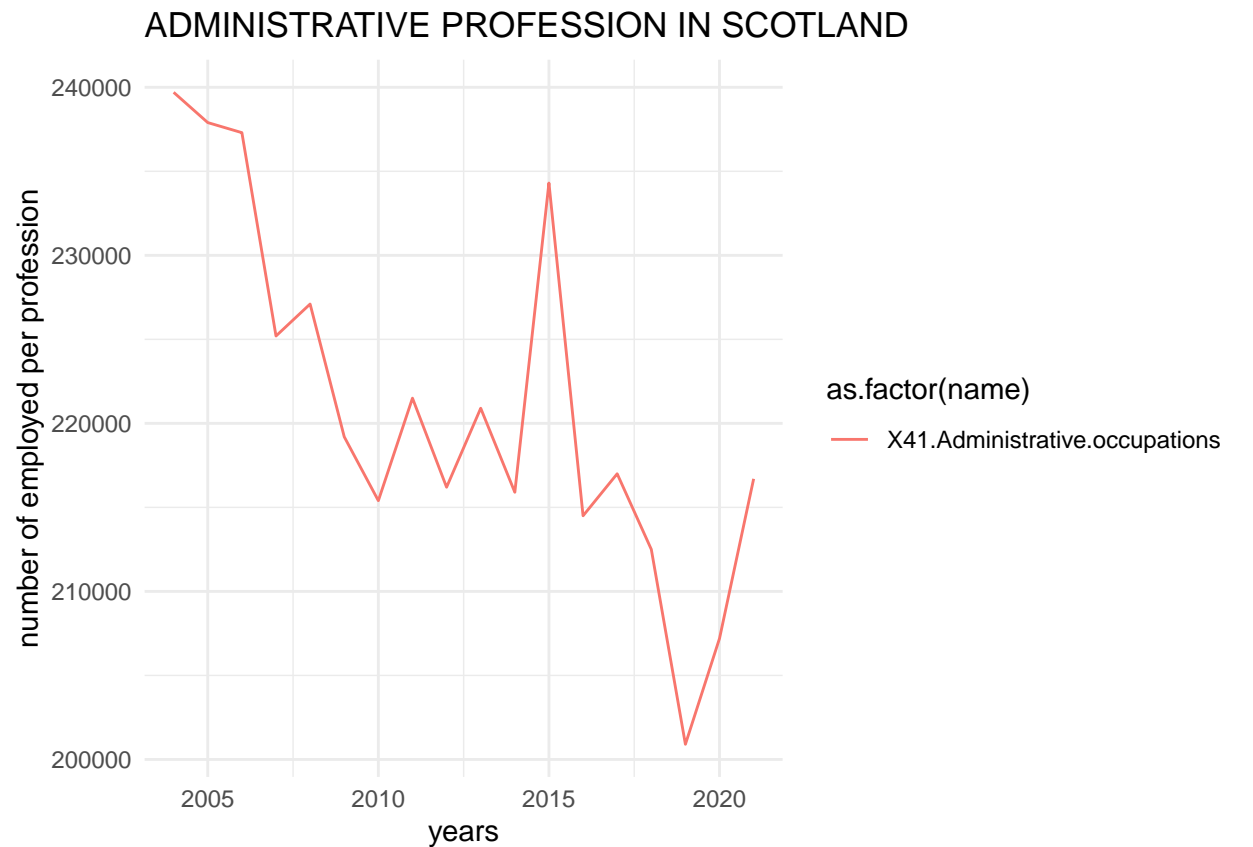


This shows there was a continual increase with effect initializing from 2019 but with significant COVID effect in 2020.

```
admin<- employ_scotland [,c("Date","Country","X41.Administrative.occupations")]
```

```
admin2 <- admin%>%
  pivot_longer(
    cols = starts_with("X")
  )
```

```
ggplot(admin2,aes(x=Date,y=value,col=as.factor(name)))+geom_line()+xlab("years")+ylab("number of employ
```

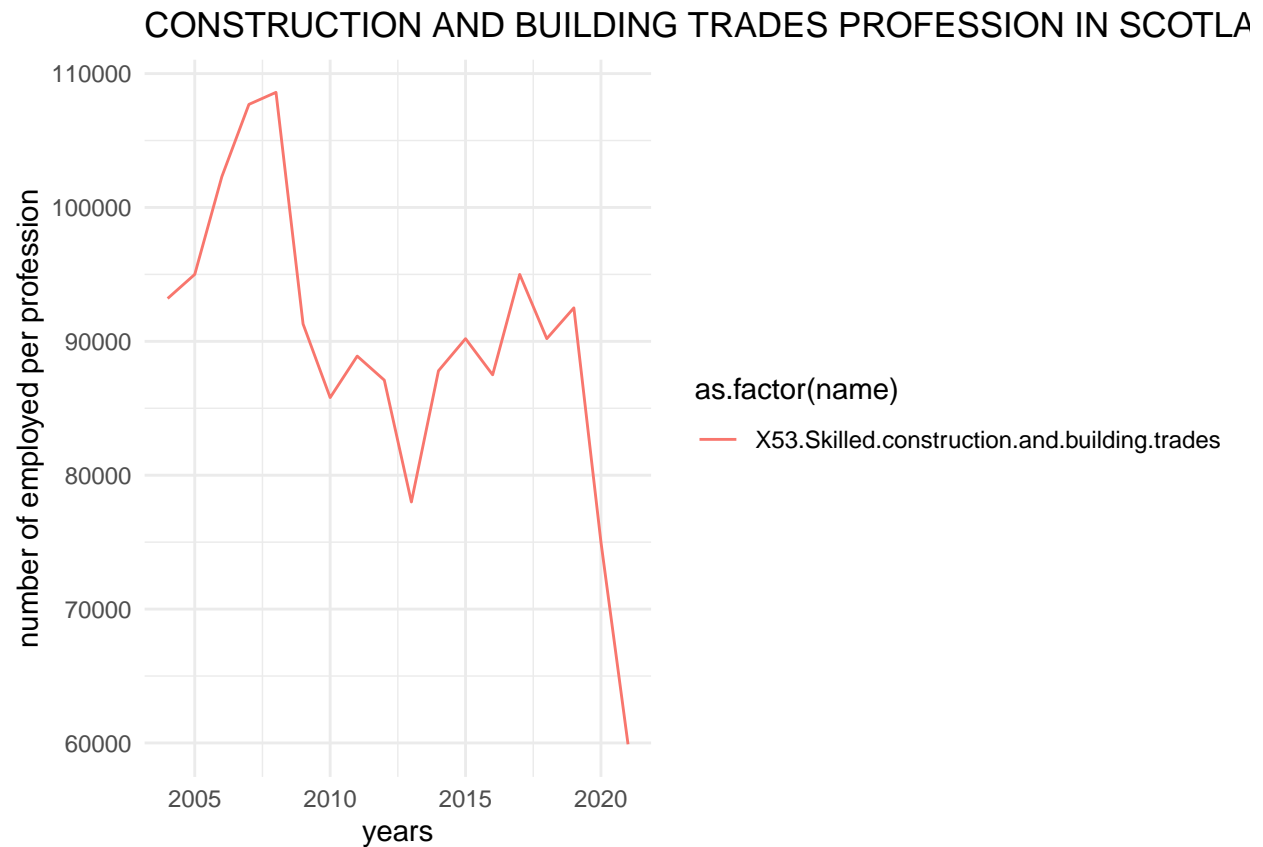


This shows there was a slight increase in 2019 but with significant COVID effect increase in 2020

```
construct<- employ_scotland[,c("Date", "Country", "X53.Skilled.construction.and.building.trades")]
```

```
construct2 <- construct%>%
  pivot_longer(
    cols = starts_with("X")
  )
```

```
ggplot(construct2,aes(x=Date,y=value,col=as.factor(name)))+geom_line()+xlab("years")+ylab("number of em
```

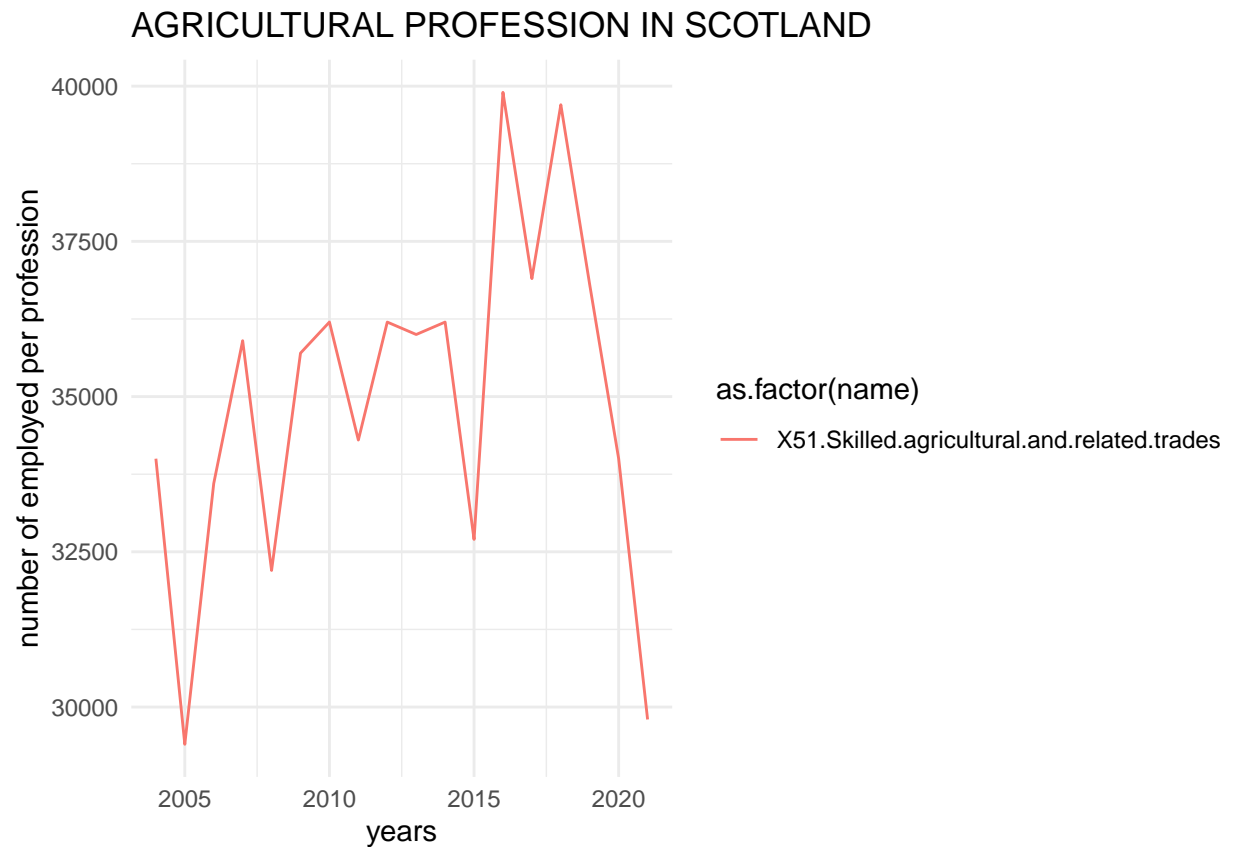


This shows there was a continual decline with effect initializing from 2019 but with significant COVID effect decrease in 2020.

```
agric<- employ_scotland[,c("Date","Country","X51.Skilled.agricultural.and.related.trades")]
```

```
agric2 <- agric%>%
  pivot_longer(
    cols = starts_with("X")
  )
```

```
ggplot(agric2,aes(x=Date,y=value,col=as.factor(name)))+geom_line()+xlab("years")+ylab("number of employ
```



This shows there was a continual decline with effect initializing from 2019 but with significant COVID effect decrease in 2020.

```
caring <- employ_scotland[,c("Date", "Country", "X61.Caring.personal.service.occupations")]
```

```
caring2 <- caring%>%
  pivot_longer(
    cols = starts_with("X")
  )
```

```
ggplot(caring2, aes(x=Date, y=value, col=as.factor(name)))+geom_line()+xlab("years")+ylab("number of employo
```

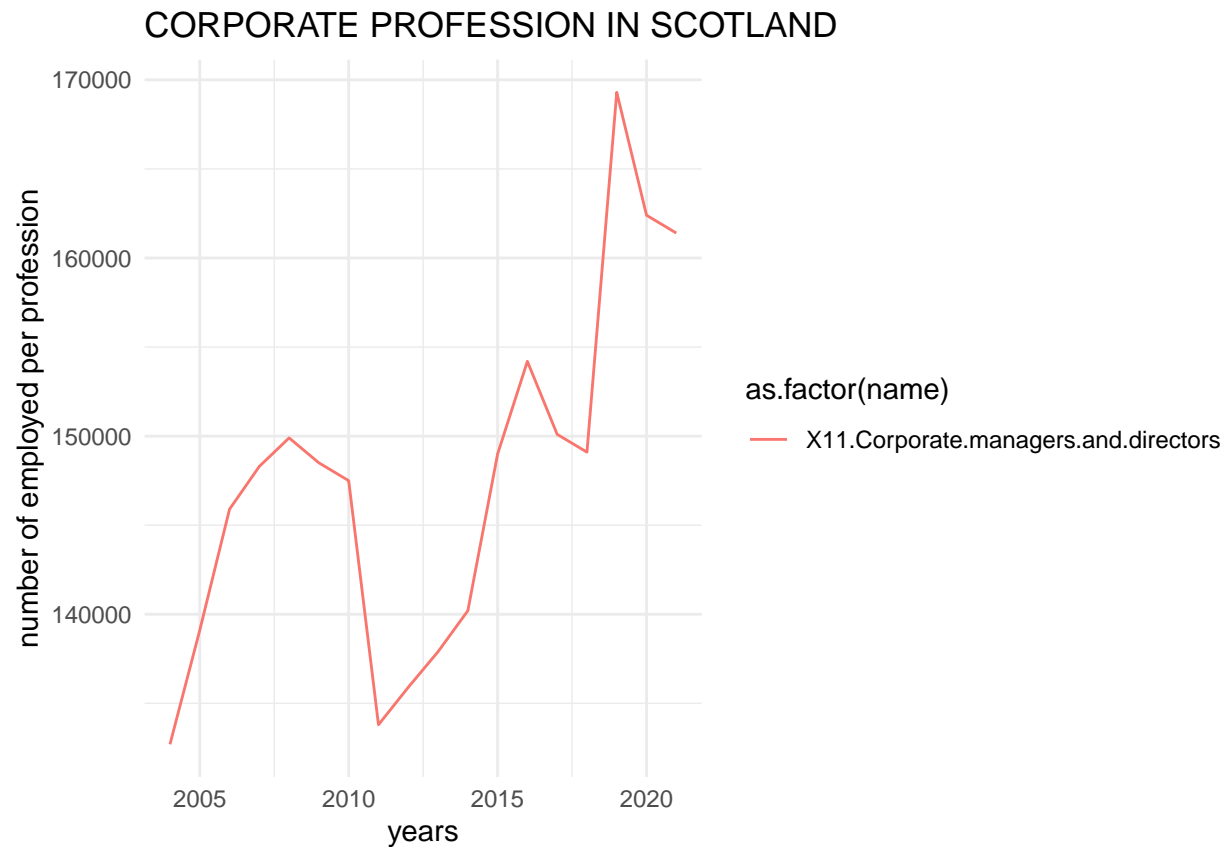


This shows there was a decline with effect initializing from 2018 but with a slight change in trend in 2020.

```
corporate <- employ_scotland[,c("Date","Country","X11.Corporate.managers.and.directors")]
```

```
corporate2 <- corporate%>%
  pivot_longer(
    cols = starts_with("X")
  )
```

```
ggplot(corporate2,aes(x=Date,y=value,col=as.factor(name)))+geom_line()+xlab("years")+ylab("number of em
```



This shows there was a decline with effect initializing from 2018 but with a slight change in trend in 2020

```
health <- employ_scotland[,c("Date", "Country", "X22.Health.professionals")]
```

```
health2 <- health%>%
  pivot_longer(
    cols = starts_with("X")
  )
```

```
ggplot(health2, aes(x=Date, y=value, col=as.factor(name)))+geom_line()+xlab("years")+ylab("number of employo
```



This shows there was a decline with effect initializing from 2018 but with almost no significance in 2020.