

Investigating Factors Influencing Academic Achievement in English Towns

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```
library(MASS)
library(tidyverse)
```

Warning: package 'tidyverse' was built under R version 4.3.2

Warning: package 'readr' was built under R version 4.3.2

Warning: package 'lubridate' was built under R version 4.3.2

```
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr      1.1.3      v readr      2.1.5
v forcats    1.0.0      v stringr    1.5.0
v ggplot2    3.4.3      v tibble     3.2.1
v lubridate  1.9.3      v tidyr      1.3.0
v purrr      1.0.2
```

```
-- Conflicts ----- tidyverse_conflicts() --
```

```
x dplyr::filter() masks stats::filter()
```

```
x dplyr::lag()     masks stats::lag()
```

```
x dplyr::select() masks MASS::select()
```

```
i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become
```

```
library(broom)
library(glm2)
```

Attaching package: 'glm2'

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

crabs

```
library(dplyr)
library(Stat2Data)
library(pROC)
```

Warning: package 'pROC' was built under R version 4.3.3

Type 'citation("pROC")' for a citation.

Attaching package: 'pROC'

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

cov, smooth, var

```
library(yardstick)
```

Attaching package: 'yardstick'

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

spec

```
library(ggplot2)
library(janitor)
```

Warning: package 'janitor' was built under R version 4.3.3

Attaching package: 'janitor'

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

```
chisq.test, fisher.test
```

```
library(here)
```

here() starts at C:/Users/zakkh/STA 210/STA210Final

```
library(fs)
library(withr)
library(lmtest)
```

Warning: package 'lmtest' was built under R version 4.3.3

Loading required package: zoo

Warning: package 'zoo' was built under R version 4.3.2

Attaching package: 'zoo'

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

```
as.Date, as.Date.numeric
```

```
#edu <- read.csv("data/english_education.csv")
edu <- readr::read_csv('https://raw.githubusercontent.com/rfordatascience/tidytuesday/master/data/2020/01/01/english_education.csv')
```

Rows: 1104 Columns: 31

-- Column specification -----

Delimiter: ","

chr (13): town11cd, town11nm, size_flag, rgn11nm, coastal, coastal_detailed,...

dbl (18): population_2011, ks4_2012_2013_counts, key_stage_2_attainment_scho...

i Use `spec()` to retrieve the full column specification for this data.

i Specify the column types or set `show_col_types = FALSE` to quiet this message.

```

working_dir <- here::here("data")

xls_path <- withr::local_tempfile(fileext = ".xlsx")
download.file(
  "https://www.ons.gov.uk/file?uri=/peoplepopulationandcommunity/educationandchildcare/data",
  xls_path,
  mode = "wb"
)

english_education <- readxl::read_xlsx(xls_path, sheet = "Data", na = "*") |>
  janitor::clean_names()

readr::write_csv(
  english_education,
  fs::path(working_dir, "english_education.csv"))

# bc <- boxcox(lm(highest_level_qualification_achieved_b_age_22_average_score ~ population
# lambda <- bc$x[which.max(bc$y)]
# english_education$population_2011_bc <- (english_education$population_2011^lambda - 1) /

# level4qual_residents35_64_2011
# activity_at_age_19_full_time_higher_education
# activity_at_age_19_apprenticeships

english_education$rgn11nm <- as.character(english_education$rgn11nm)

english_education$rgn11nm_combined <- "Other"

#Combining regions where it makes sense geographically
english_education$rgn11nm_combined[english_education$rgn11nm %in% c("South East", "South W
english_education$rgn11nm_combined[english_education$rgn11nm == "North East"] <- "North Ea
english_education$rgn11nm_combined[english_education$rgn11nm == "North West"] <- "North We

english_education$rgn11nm_combined <- factor(english_education$rgn11nm_combined)
english_education$rgn11nm_combined <- factor(english_education$rgn11nm_combined)

#Baseline
english_education$rgn11nm_combined <- relevel(english_education$rgn11nm_combined, ref = "S

#Final model that includes an interaction term
m1 <- lm(

```

```

highest_level_qualification_achieved_b_age_22_average_score ~
  level4qual_residents35_64_2011*level_3_at_age_18 +
  activity_at_age_19_full_time_higher_education +
  level4qual_residents35_64_2011 +
  rgn11nm_combined +
  level_3_at_age_18 +
  key_stage_4_attainment_school_year_2012_to_2013,
data = english_education
)
summary(m1)

```

Call:

```

lm(formula = highest_level_qualification_achieved_b_age_22_average_score ~
  level4qual_residents35_64_2011 * level_3_at_age_18 + activity_at_age_19_full_time_higher_education +
  level4qual_residents35_64_2011 + rgn11nm_combined + level_3_at_age_18 +
  key_stage_4_attainment_school_year_2012_to_2013, data = english_education)

```

Residuals:

	Min	1Q	Median	3Q	Max
	-0.38428	-0.05926	-0.00773	0.05484	0.38699

Coefficients:

	Estimate	Std. Error
(Intercept)	1.8241957	0.0737405
level4qual_residents35_64_2011Low	0.2471406	0.0764781
level4qual_residents35_64_2011Medium	0.2875048	0.0783453
level_3_at_age_18	0.0144596	0.0012259
activity_at_age_19_full_time_higher_education	0.0200731	0.0006169
rgn11nm_combinedNorth East	0.0918015	0.0129024
rgn11nm_combinedNorth West	0.0571483	0.0093396
rgn11nm_combinedOther	0.0323596	0.0071201
key_stage_4_attainment_school_year_2012_to_2013	0.0031308	0.0005040
level4qual_residents35_64_2011Low:level_3_at_age_18	-0.0033080	0.0012279
level4qual_residents35_64_2011Medium:level_3_at_age_18	-0.0040273	0.0012151

	t value	Pr(> t)
(Intercept)	24.738	< 2e-16 ***
level4qual_residents35_64_2011Low	3.232	0.001268 **
level4qual_residents35_64_2011Medium	3.670	0.000255 ***
level_3_at_age_18	11.795	< 2e-16 ***
activity_at_age_19_full_time_higher_education	32.538	< 2e-16 ***
rgn11nm_combinedNorth East	7.115	2.03e-12 ***

rgn11nm_combinedNorth West	6.119	1.31e-09	***
rgn11nm_combinedOther	4.545	6.11e-06	***
key_stage_4_attainment_school_year_2012_to_2013	6.212	7.42e-10	***
level4qual_residents35_64_2011Low:level_3_at_age_18	-2.694	0.007170	**
level4qual_residents35_64_2011Medium:level_3_at_age_18	-3.314	0.000949	***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.09655 on 1088 degrees of freedom

(5 observations deleted due to missingness)

Multiple R-squared: 0.9282, Adjusted R-squared: 0.9275

F-statistic: 1406 on 10 and 1088 DF, p-value: < 2.2e-16

```
#transforming to log odds
#english_education <- english_education %>%
# mutate(log_odds_ks4asy = log(key_stage_4_attainment_school_year_2012_to_2013 / (1 #- ke

#m1 <- lm(education_score ~ population_2011, data = english_education)

#summary(m1)

#m1 <- lm(education_score ~ factor(size_flag), data = english_education)

#summary(m1)

#better option - population_2011
#m1 <- lm(education_score ~ factor(size_flag)+factor(university_flag)+factor###(job_densi
#summary(m1)

#english_education <- na.omit(english_education) #complete case analysis for all variables

english_education$rgn11nm <- as.character(english_education$rgn11nm)

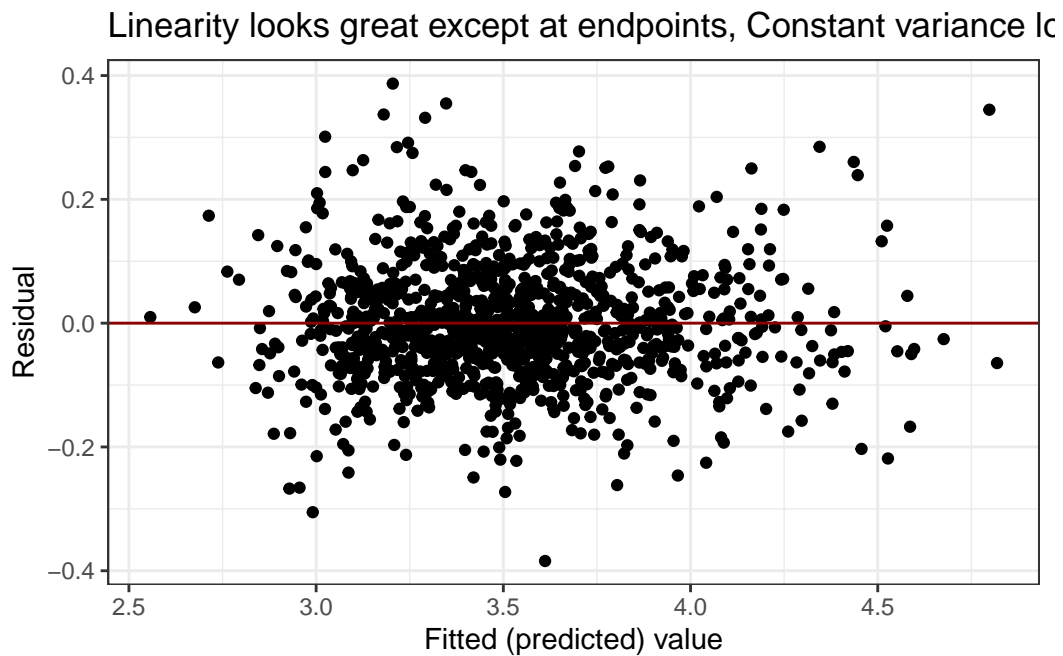
english_education$rgn11nm_combined <- "Other"

english_education$rgn11nm_combined[english_education$rgn11nm %in% c("South East", "South W
english_education$rgn11nm_combined[english_education$rgn11nm == "North East"] <- "North Ea
english_education$rgn11nm_combined[english_education$rgn11nm == "North West"] <- "North We

english_education$rgn11nm_combined <- factor(english_education$rgn11nm_combined)
english_education$rgn11nm_combined <- factor(english_education$rgn11nm_combined)
```

```
english_education$rgn11nm_combined <- relevel(english_education$rgn11nm_combined, ref = "S
```

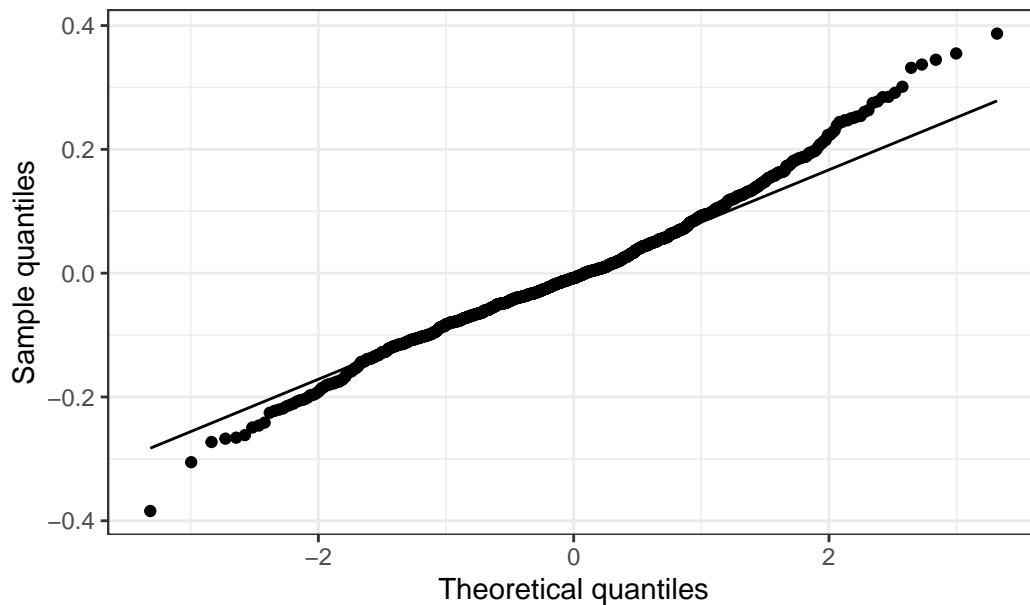
```
#Residual Plot  
ggplot(m1, aes(x = .fitted, y = .resid)) +  
  geom_point() +  
  geom_hline(yintercept = 0, color = "darkred") +  
  labs(x = "Fitted (predicted) value", y = "Residual", title = "Linearity looks great exce  
  theme_bw()
```



```
m1_aug <- augment(m1)
```

```
#QQ Plot  
ggplot(m1, aes(sample = .resid)) +  
  stat_qq() +  
  stat_qq_line() +  
  theme_bw() +  
  labs(x = "Theoretical quantiles",  
       y = "Sample quantiles", title = "Slight deviations but very reasonably normal")
```

Slight deviations but very reasonably normal



```
#Our custom method of assessing constant variance - splitting into even intervals, calculating  
quantiles <- quantile(m1_aug$.fitted, probs = seq(0, 1, by = 0.2))
```

```
variance_intervals_df <- data.frame(interval = character(0), variance = numeric(0))
```

```
for (i in 1:(length(quantiles) - 1)) {
```

```
  subset_data <- m1_aug %>%
```

```
    filter(.fitted >= quantiles[i] & .fitted < quantiles[i + 1])
```

```
  interval_name <- paste(round(quantiles[i], 2), "-", round(quantiles[i + 1], 2), sep="")
```

```
  variance_value <- var(subset_data$.resid)
```

```
  variance_intervals_df <- rbind(variance_intervals_df,  
                                data.frame(interval=interval_name,  
                                             variance=variance_value))
```

```
}
```

```
print(variance_intervals_df)
```

	interval	variance
1	2.56-3.22	0.010439236
2	3.22-3.4	0.008603479


```
3 3.4-3.56 0.007706760
4 3.56-3.78 0.009201645
5 3.78-4.82 0.010143797
```

```
#Formal test for constant variance
bptest_result <- bptest(m1)

print(bptest_result)
```

studentized Breusch-Pagan test

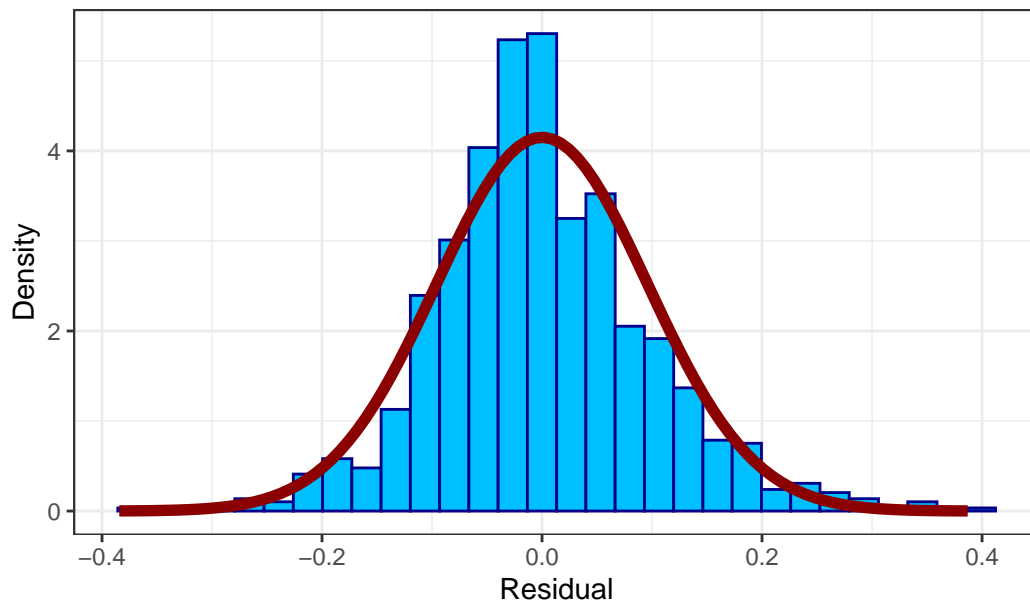
```
data: m1
BP = 27.512, df = 10, p-value = 0.00216
```

```
resid_mean <- mean(m1_aug$.resid, na.rm = TRUE)
resid_sd <- sd(m1_aug$.resid, na.rm = TRUE)

#Histogram compared to normal distribution
ggplot(m1_aug, aes(x = .resid)) +
  geom_histogram(aes(y = ..density..),
                 fill = "deepskyblue", color = "darkblue", bins = 30) +
  stat_function(fun = dnorm,
               args = list(mean = resid_mean, sd = resid_sd),
               color = "darkred", lwd = 2) +
  labs(x = "Residual", y = "Density", title = "Symmetric tails and great fit to normal dis") +
  theme_bw()
```

Warning: The dot-dot notation (`..density..`) was deprecated in ggplot2 3.4.0.
i Please use `after_stat(density)` instead.

Symmetric tails and great fit to normal distribution



```
#transforming back
#predicted_log_odds <- predict(m1, type = "response")
#predicted_proportions <- exp(predicted_log_odds) / (1 + exp(predicted_log_odds))

#Finding min and max values taken on by variables, columns of the dataset.
min_max_education_score <- english_education %>%
  summarise(min_education_score = min(education_score, na.rm = TRUE),
            max_education_score = max(education_score, na.rm = TRUE))

min_max_population_2011 <- english_education %>%
  summarise(min_population_2011 = min(population_2011, na.rm = TRUE),
            max_population_2011 = max(population_2011, na.rm = TRUE))

min_max_highest_qualification <- english_education %>%
  summarise(min_highest_qualification = min(highest_level_qualification_achieved_b_age_22, na.rm = TRUE),
            max_highest_qualification = max(highest_level_qualification_achieved_b_age_22, na.rm = TRUE))

min_max_level_3_age_18 <- english_education %>%
  summarise(min_level_3_age_18 = min(level_3_at_age_18, na.rm = TRUE),
            max_level_3_age_18 = max(level_3_at_age_18, na.rm = TRUE))
```

```

min_max_activity_age_19 <- english_education %>%
  summarise(min_activity_age_19 = min(activity_at_age_19_employment_with_earnings_above_10
    max_activity_age_19 = max(activity_at_age_19_employment_with_earnings_above_10

min_max_key_stage_4 <- english_education %>%
  summarise(min_key_stage_4 = min(key_stage_4_attainment_school_year_2012_to_2013, na.rm =
    max_key_stage_4 = max(key_stage_4_attainment_school_year_2012_to_2013, na.rm =

min_max_df <- bind_rows(
  min_max_education_score,
  min_max_population_2011,
  min_max_highest_qualification,
  min_max_level_3_age_18,
  min_max_activity_age_19,
  min_max_key_stage_4
)

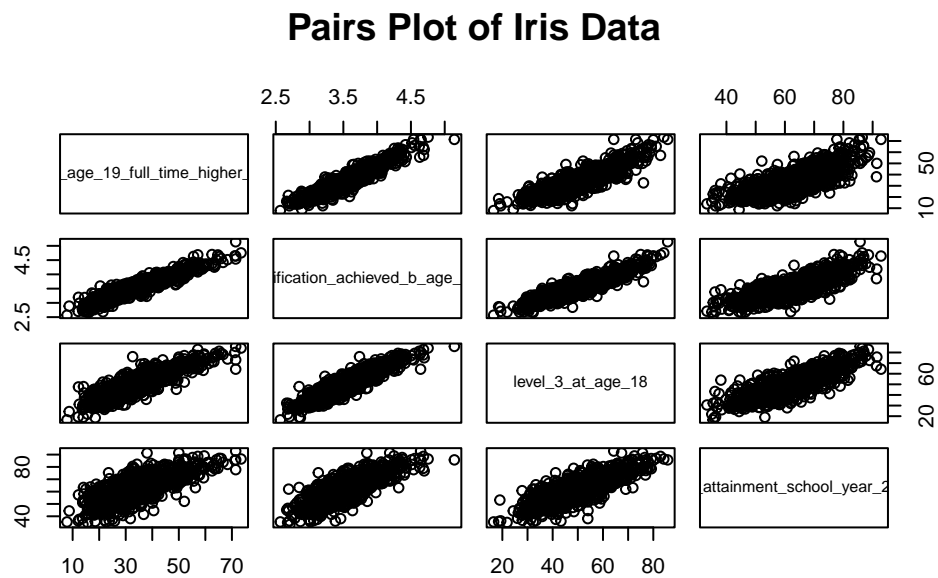
print(min_max_df)

# A tibble: 6 x 12
  min_education_score max_education_score min_population_2011
      <dbl>             <dbl>             <dbl>
1      -10.0           11.9              NA
2         NA           NA              5003
3         NA           NA              NA
4         NA           NA              NA
5         NA           NA              NA
6         NA           NA              NA
# i 9 more variables: max_population_2011 <dbl>,
#   min_highest_qualification <dbl>, max_highest_qualification <dbl>,
#   min_level_3_age_18 <dbl>, max_level_3_age_18 <dbl>,
#   min_activity_age_19 <dbl>, max_activity_age_19 <dbl>,
#   min_key_stage_4 <dbl>, max_key_stage_4 <dbl>

edu_pairs <- english_education |>
  select(
    activity_at_age_19_full_time_higher_education,
    highest_level_qualification_achieved_b_age_22_average_score,
    level_3_at_age_18,
    key_stage_4_attainment_school_year_2012_to_2013) |>
  mutate()

```

```
pairs(edu_pairs[, 1:4], main = "Pairs Plot of Iris Data")
```



```
ggplot(english_education, aes(x = factor(rgn11nm_combined))) +
  geom_bar() +
  labs(x = "Region (rgn11nm_combined)", y = "Count", title = "Relative Distributions of Ne
```

