

From Deprivation to Degree

Exploring the Impact of Local Income Levels on Youth Education in England

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1 Executive summary

This report investigates **how town-level income deprivation affects educational attainment among English youth aged 18 to 22**. Using national data, we analyze academic outcomes at three life stages: Level 3 qualification at 18, university entry at 19, and Level 6+ attainment by 22. Results show a clear **upward trend**—youth from lower deprivation towns consistently outperform those from higher deprivation areas, especially by age 22.

These findings highlight structural inequalities and suggest the need for targeted educational support in deprived communities.

2 Introduction

Regional inequality in education remains a pressing issue in England, particularly among young people transitioning from secondary to higher education (The Times, 2024). This study examines **whether income deprivation at the town level impacts academic outcomes for individuals aged 18 to 22**. We focus on three key milestones: Level 3 qualifications at age 18, full-time university participation at age 19, and Level 6 or higher qualification by age 22. These stages reflect national benchmarks for educational progression.

To maintain analytical consistency, cities are excluded due to their distinctive socio-demographic structures, such as dense populations and diverse education systems. We also remove entries with missing or unclear income classifications to ensure accurate results. The dataset is grouped by income level and town size, and analyzed using both summary statistics and visualizations. Trends are observed across large, medium, and small towns.

Findings reveal a consistent association: **as income deprivation decreases, educational attainment increases**. This analysis deepens our understanding of structural disadvantage and supports the development of more equitable, data-driven education policies.

3 Research Question

This study will focus on:

Does the income level of a town affect the educational outcomes of young people?

4 Data Loading

In this study, our group chose to **download the dataset locally** rather than relying on an online dataset. This approach **improves the reproducibility of the analysis** because local data ensures continuous access to a stable version of the data and avoids remote sources updating or deleting the data resulting in different analysis results. Also, **local datasets are more stable** as they do not rely on a network connection, which avoids interruptions due to application program interface limitations or network problems. It also improves processing speed and supports offline analysis, which is essential for **a reliable and efficient research workflow**.

We use the `english_education.csv` dataset downloading from the [TidyTuesday](#), originally sourced from [The UK Office for National Statistics \(ONS\)](#). The dataset contains education and deprivation data for over 1,000 English towns from 2012 to 2013 school year, offering town-level observations across key variables such as `income_flag`, `size_flag`, `level_3_at_age_18`, `activity_at_age_19_full_time_higher_education`, and `highest_level_qualification_achieved_by_age_22_level_6_or_above`. These indicators provide a robust basis for analyzing **educational progression** in relation to **local income deprivation**.

```
edu_df <- read_csv("../data/english_education.csv")

selected_vars <- c(
  "income_flag",
  "level_3_at_age_18",
  "activity_at_age_19_full_time_higher_education",
  "highest_level_qualification_achieved_by_age_22_level_6_or_above",
  "size_flag"
)

edu_df_selected <- edu_df %>%
  select(all_of(selected_vars))%>%
  rename(
    level3_at_18 = level_3_at_age_18,
    uni_at_19 = activity_at_age_19_full_time_higher_education,
    level6_at_22 = highest_level_qualification_achieved_by_age_22_level_6_or_above
  )
```

5 Dataset introduction

5.1 Variables Description

From Table 1, the variables listed here represent a selected subset of those most relevant to our analysis. The dataset is suitable for **examining how structural income deprivation may affect academic progression among young people**. The original data has been archived in the project repository and here is [the location of the data](#).

```
edu_variable_desc <- data.frame(
  Variable = c(
    "income_flag",
    "level3_at_18",
    "uni_at_19",
    "level6_at_22",
    "size_flag"
  ),
  Description = c(
    "Categorical income deprivation classification of the town (lower/mid/higher). Used
    ↪ as the independent grouping variable.",
    "Proportion of young people in the 2012/13 cohort who achieved Level 3 qualifications
    ↪ at age 18.",
    "Proportion of the cohort who entered full-time higher education by age 19.",
    "Proportion of the cohort who attained Level 6 or above qualifications (e.g.,
    ↪ Bachelor's degree) by age 22.",
    "Town/city size classification based on 2011 census data (used to identify and
    ↪ exclude cities)."
  )
)

kable(
  edu_variable_desc,
  col.names = c("Variable Names", "Variable Description"),
  booktabs = TRUE,
  longtable = TRUE) %>%
  kable_styling(
    latex_options = c("striped", "hold_position", "scale_down"),
    stripe_color = "gray!15",
    font_size = 10
  ) %>%
```

```
column_spec(1, width = "3.5cm") %>%
column_spec(2, width = "11cm")
```

Table 1: Key Variables Used in the Analysis

Variable Names	Variable Description
income_flag	Categorical income deprivation classification of the town (lower/mid/higher). Used as the independent grouping variable.
level3_at_18	Proportion of young people in the 2012/13 cohort who achieved Level 3 qualifications at age 18.
uni_at_19	Proportion of the cohort who entered full-time higher education by age 19.
level6_at_22	Proportion of the cohort who attained Level 6 or above qualifications (e.g., Bachelor's degree) by age 22.
size_flag	Town/city size classification based on 2011 census data (used to identify and exclude cities).

5.2 Dataset Description

The dataset contains a total of 1104 observations and 31 variables.

The dataset contains a total of `nrow(edu_df)` observations and `ncol(edu_df)` variables.

Figure 1: Inline R code

See Figure 1 for the detail of the *Inline R code* used.

Table 2: First 2 rows of the data

```
head(edu_df_selected, 2)
```

```
# A tibble: 2 x 5
  income_flag level3_at_18 uni_at_19 level6_at_22 size_flag
  <chr>          <dbl>    <dbl>    <dbl> <chr>
1 Higher deprivation towns    50.8    30.8    NA Small Towns
2 Mid deprivation towns      60.1    41.8   33.3 Small Towns
```

See Table 2 for the detail of the first 2 rows of the data and the types of variables in the dataset.

6 Methodology

```
edu_df_selected %>%
  count(income_flag) %>%
  mutate(percentage = round(n / sum(n) * 100, 1)) %>%
  kable(
    col.names = c("Income Flag", "Count", "Percentage (%)"),
    digits = 1,
    booktabs = TRUE) %>%
  kable_styling(
    latex_options = c("striped", "hold_position"),
    stripe_color = "gray!15",
    font_size = 10
  )
```

Table 3: Proportion of Each Income Flag Category

Income Flag	Count	Percentage (%)
Cities	18	1.6
Higher deprivation towns	433	39.2
Lower deprivation towns	444	40.2
Mid deprivation towns	205	18.6
NA	4	0.4

While reviewing the data (Table 3), our group discover that there is **missing data** (NA value) in the income categories `income_flag`. In order **not to affect the accuracy of the subsequent analysis**, this missing data will not be included in the analysis. Also, since the sample in `Cities` of `income_flag` is **too small and less representative**, it is removed in the subsequent analysis.

```
edu_df_clean <- edu_df_selected %>%
  filter(!is.na(income_flag) & income_flag != "Cities") %>%
  mutate(income_flag = factor(income_flag,
                              levels = c("Higher deprivation towns",
```

```

"Mid deprivation towns",
"Lower deprivation towns"))))

```

```

table1 <- edu_df_clean %>%
  group_by(income_flag) %>%
  summarise(
    mean_level3 = mean(level3_at_18, na.rm = TRUE),
    sd_level3 = sd(level3_at_18, na.rm = TRUE),
    mean_uni19 = mean(uni_at_19, na.rm = TRUE),
    sd_uni19 = sd(uni_at_19, na.rm = TRUE),
    mean_level6 = mean(level6_at_22, na.rm = TRUE),
    sd_level6 = sd(level6_at_22, na.rm = TRUE)
  ) %>%
  rename(
    "Income Group" = income_flag,
    "Mean: Level 3 at 18" = mean_level3,
    "SD: Level 3 at 18" = sd_level3,
    "Mean: HE at 19" = mean_uni19,
    "SD: HE at 19" = sd_uni19,
    "Mean: Level 6 at 22" = mean_level6,
    "SD: Level 6 at 22" = sd_level6
  )

kable(table1,
      digits = 2,
      booktabs = TRUE,
      longtable = TRUE
) %>%
  kable_styling(
    latex_options = c("striped", "hold_position",
                      "scale_down", "repeat_header", "longtable"),
    stripe_color = "gray!15",
    font_size = 9
  ) %>%
  column_spec(1, width = "2cm") %>%
  column_spec(2, width = "1.8cm") %>%
  column_spec(3, width = "1.8cm") %>%
  column_spec(4, width = "1.8cm") %>%
  column_spec(5, width = "1.8cm") %>%

```



```
column_spec(6, width = "1.8cm") %>%
column_spec(7, width = "1.8cm")
```

Table 4: Mean and SD of Educational Outcomes by Income Group(Cities Excluded)

Income Group	Mean: Level 3 at 18	SD: Level 3 at 18	Mean: HE at 19	SD: HE at 19	Mean: Level 6 at 22	SD: Level 6 at 22
Higher deprivation towns	41.59	7.35	27.13	6.69	23.99	4.92
Mid deprivation towns	47.82	7.59	31.32	7.20	27.99	5.69
Lower deprivation towns	57.96	9.68	40.78	10.33	35.68	8.33

Table 4 shows that **education outcomes improve significantly at all measured stages as income levels rise**.

For instance, the mean proportion of individuals achieving Level 3 qualifications by age 18 rises from **41.59%** in **higher deprivation** towns to **57.96%** in **lower deprivation** towns. Similar upward trends are observed for higher education participation at age 19 (**27.13%** to **40.78%**) and for achieving Level 6 qualifications by age 22 (**23.99%** to **35.68%**).

The standard deviations are relatively consistent across groups. This suggests that while variation exists, the overall improvement in outcomes is not due to a small number of high performers.

```
edu_long <- edu_df_clean %>%
  pivot_longer(
    cols = c(level3_at_18,
              uni_at_19,
              level6_at_22),
    names_to = "education_stage",
    values_to = "value"
  )
```

```

edu_long$income_flag <- recode(edu_long$income_flag,
  "Higher deprivation towns" = "Higher\ndeprivation\ntowns",
  "Mid deprivation towns" = "Mid\ndeprivation\ntowns",
  "Lower deprivation towns" = "Lower\ndeprivation\ntowns"
)

edu_long$education_stage <- recode(edu_long$education_stage,
  "level3_at_18" = "Level 3 at 18",
  "uni_at_19" = "HE at 19",
  "level6_at_22" = "Level 6 at 22"
)

ggplot(edu_long, aes(x = income_flag, y = value, fill = income_flag)) +
  geom_boxplot() +
  facet_wrap(~education_stage, scales = "free_y") +
  labs(x = "Income Group", y = "Proportion") +
  theme_minimal() +
  theme(axis.text.x = element_text(size = 9, lineheight = 1.1))

```

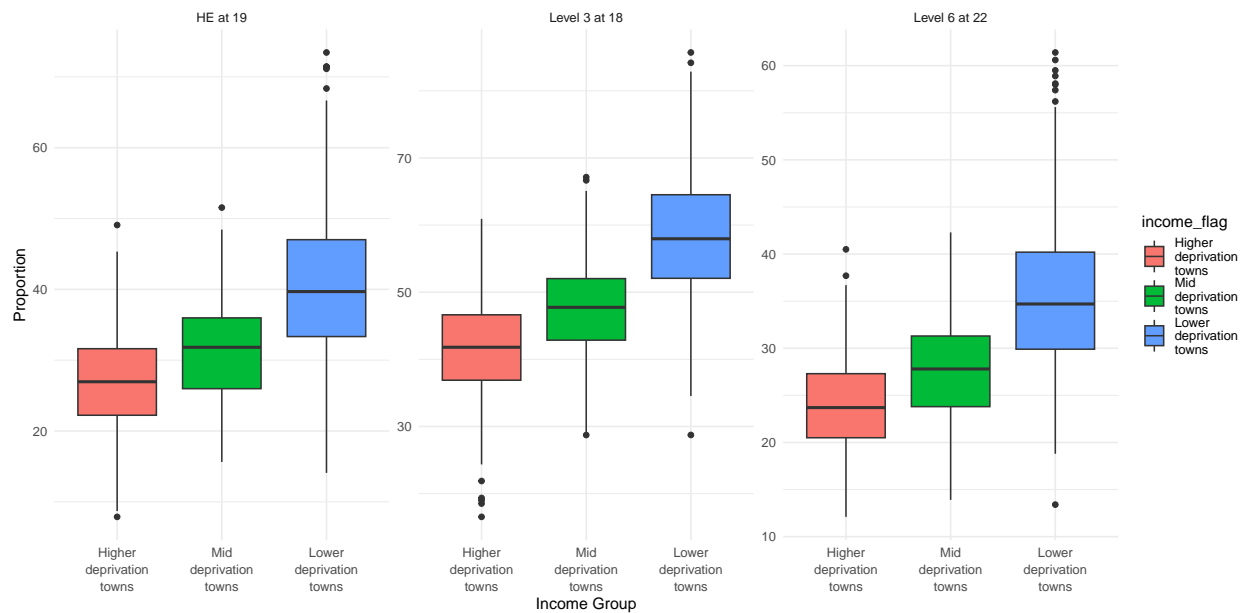


Figure 2: Educational Outcomes by Income Group(Cities Excluded)

Figure 2 illustrates a clear upward trend in educational attainment as town income levels increase. In all three stages, young people in lower deprivation towns consistently outperform their peers in higher deprivation towns.

Specifically, the **median** proportions of youth achieving Level 3 qualifications at age 18, participating in full-time higher education at age 19, and attaining Level 6 qualifications by age 22 are **all higher in towns with lower deprivation**. The **interquartile ranges** are also **narrower** in **lower deprivation** towns, suggesting more consistent performance within these areas.

In contrast, **higher deprivation** towns exhibit not only **lower median outcomes** but also **greater variability**, as reflected by the wider spread of the boxplots and more outliers.

7 Result

```
summary_df <- edu_df_clean %>%
  group_by(income_flag, size_flag) %>%
  summarise(
    level3 = mean(level3_at_18, na.rm = TRUE),
    uni19 = mean(uni_at_19, na.rm = TRUE),
    level6 = mean(level6_at_22, na.rm = TRUE)
  ) %>%
  pivot_longer(
    cols = c(level3, uni19, level6),
    names_to = "education_stage",
    values_to = "mean_value"
  )

summary_df$education_stage <- recode(summary_df$education_stage,
  "level3" = "Level 3 at Age 18",
  "uni19" = "University at 19",
  "level6" = "Level 6+ by Age 22"
)

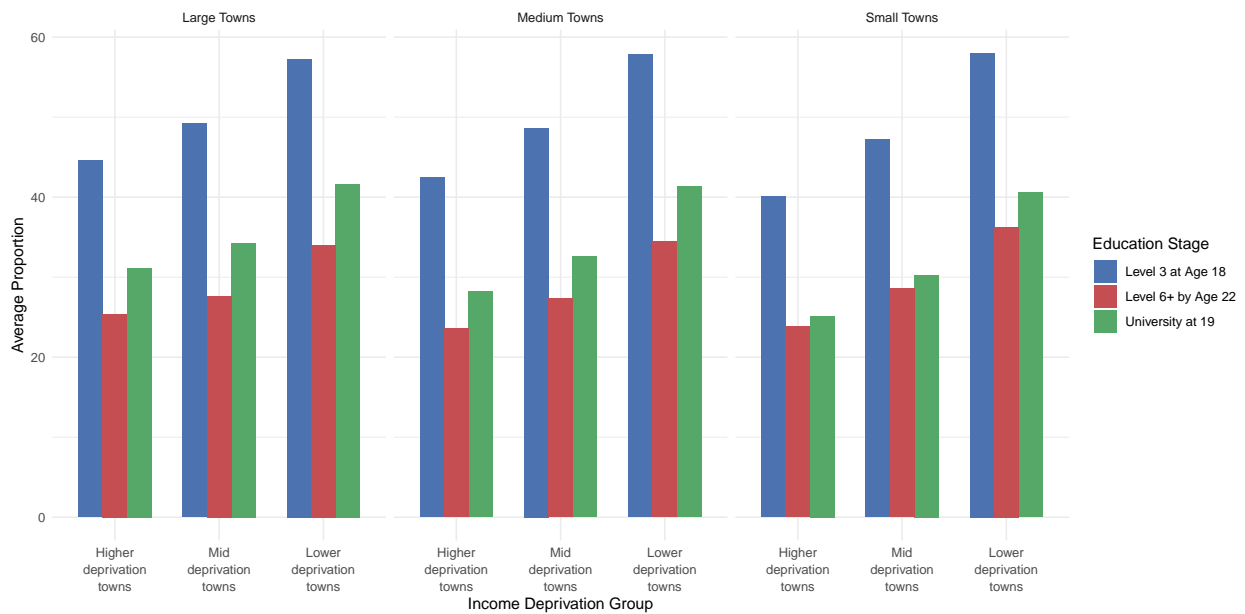
summary_df$income_flag <- recode(summary_df$income_flag,
  "Higher deprivation towns" = "Higher\ndeprivation\ntowns",
  "Mid deprivation towns" = "Mid\ndeprivation\ntowns",
  "Lower deprivation towns" = "Lower\ndeprivation\ntowns"
)

ggplot(summary_df, aes(x = income_flag, y = mean_value, fill = education_stage)) +
  geom_col(position = "dodge", width = 0.7) +
  facet_wrap(~size_flag) +
  labs(
```

```

x = "Income Deprivation Group",
y = "Average Proportion",
fill = "Education Stage"
) +
scale_fill_manual(
  values = c(
    "Level 3 at Age 18" = "#4C72B0",
    "University at 19" = "#55A868",
    "Level 6+ by Age 22" = "#C44E52"
  )
) +
theme_minimal() +
theme(axis.text.x = element_text(size = 9, lineheight = 1.1))

```



(a) Comparison of educational outcomes across income levels and town sizes

Figure 3: Educational Attainment by Income Deprivation Level and Town Size (Cities Excluded)

As shown in Figure 3, **educational attainment increases consistently as income deprivation decreases, across all town sizes**. Towns with **lower deprivation levels** demonstrate the **highest average proportions** of students achieving Level 3 qualifications at age 18, entering university at age 19, and obtaining Level 6 or higher qualifications by age 22.

This pattern holds for large, medium, and small towns, though the gap between high and low deprivation groups appears most pronounced in small towns, particularly at the Level 6+ stage.

In small towns, the difference in university attainment between high and low deprivation groups exceeds 10 percentage points.

These results **reinforce a strong association between local income levels and youth educational outcomes**, and suggest that deprivation may have an even stronger impact in smaller towns.

8 Discussion

Our analysis reveals a consistent trend: towns with lower levels of income deprivation demonstrate stronger educational outcomes at ages 18, 19, and 22. This association holds across all town sizes and is particularly pronounced in smaller towns, where the disparity in Level 6 qualifications is most significant. These findings align with recent research by the UK Office for National Statistics (ONS), which suggests that less deprived areas may offer more supportive conditions for academic progression.

Nonetheless, several limitations of this study must be acknowledged. First, as an observational analysis, our findings illustrate correlation rather than causation. While the relationship between deprivation and lower attainment is strong, we cannot definitively claim that income deprivation directly causes poorer educational outcomes. Second, the use of broad categorical income groupings may mask finer-grained local differences, potentially obscuring key contextual factors. Third, we deliberately excluded cities to ensure analytical consistency; however, this decision limits the generalisability of our findings, particularly to urban populations who often face distinct educational barriers. Lastly, the dataset is limited to a single cohort of students from the 2012/13 academic year, meaning it may not capture evolving socioeconomic dynamics or long-term trends in educational performance.

Despite these constraints, the findings clearly highlight structural inequalities that disadvantage youth in more deprived towns. This underscores the need for targeted, evidence-based policy interventions to reduce educational disparities. Future research should incorporate longitudinal data and more granular socioeconomic indicators to deepen our understanding of these patterns and inform more effective policy responses.

9 Conclusion

This study found a clear relationship between **income deprivation** and **young people’s educational attainment** at town level in England. At all three stages of education, young people from **less deprived towns** and cities consistently **outperformed** those from more deprived areas. This

trend exists across towns and cities of all sizes, and is particularly evident in **smaller towns**. While the findings are consistent with national statistics and highlight structural inequalities, **caution should be exercised in interpreting the results** due to the observational nature and limited scope of the study.

10 Recommendations

To address the disparities observed, education policy should **prioritize support for students in high-deprivation towns**. This may include:

- Enhance educational funding and targeted support in high-deprivation towns
- Expand access programs to encourage university progression
- Better access to higher education pathways
- Integrate deprivation indicators into education planning and policy to ensure equitable resource distribution

In addition, **future research** should expand the study's dataset. There is a need to include **multiple clusters and urban areas** and to **integrate qualitative indicators** such as school quality and student support services as study variables. These efforts will contribute to **a more comprehensive understanding of how income poverty affects educational trajectories** and inform the development of more equitable education strategies.

11 Citation

R and R Package

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Website

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