

analysis.R

byrds

2025-05-07

```
# Run "./data/new_data97-educational-data/new_data97-educational-data.R" first.  
# This file runs analyses on two rounds of the NLSY97 data set.  
setwd("C:/Users/byrds/Documents/rStudio_work/social_research")
```

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

```
## Warning: package 'ggplot2' was built under R version 4.4.3
```

```
## Warning: package 'purrr' was built under R version 4.4.3
```

```
## Warning: package 'lubridate' was built under R version 4.4.3
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --  
## v dplyr      1.1.4      v readr      2.1.5  
## v forcats    1.0.0      v stringr   1.5.1  
## v ggplot2    3.5.2      v tibble    3.2.1  
## v lubridate  1.9.4      v tidyr     1.3.1  
## v purrr      1.0.4
```

```
## -- 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 errors
```

```
library(extrafont)
```

```
## Registering fonts with R
```

```
library(ggthemes)
```

```
## Warning: package 'ggthemes' was built under R version 4.4.2
```

```
library(hrbrthemes)  
library(thematic)
```

```
## Warning: package 'thematic' was built under R version 4.4.2
```

```
library(colorspace)
library(addinslist)
library(gmodels)
```

```
## Warning: package 'gmodels' was built under R version 4.4.2
```

```
library(RColorBrewer)
library(DescTools)
```

```
## Warning: package 'DescTools' was built under R version 4.4.3
```

```
## Registered S3 method overwritten by 'DescTools':
##   method          from
##   reorder.factor  gdata
```

```
library(viridis)
```

```
## Loading required package: viridisLite
```

```
library(ggpmisc)
```

```
## Warning: package 'ggpmisc' was built under R version 4.4.2
```

```
## Loading required package: ggpp
```

```
## Warning: package 'ggpp' was built under R version 4.4.2
```

```
## Registered S3 methods overwritten by 'ggpp':
##   method          from
##   heightDetails.titleGrob ggplot2
##   widthDetails.titleGrob  ggplot2
##
## Attaching package: 'ggpp'
##
## The following object is masked from 'package:ggplot2':
##
##   annotate
```

```
library(naniar)
```

```
## Warning: package 'naniar' was built under R version 4.4.3
```

```
library(broom)
```

```
## Warning: package 'broom' was built under R version 4.4.3
```

```
library(mice)
```

```
## Warning: package 'mice' was built under R version 4.4.3
```

```
##  
## Attaching package: 'mice'  
##  
## The following object is masked from 'package:stats':  
##  
##     filter  
##  
## The following objects are masked from 'package:base':  
##  
##     cbind, rbind
```

```
library(survey)
```

```
## Warning: package 'survey' was built under R version 4.4.3
```

```
## Loading required package: grid  
## Loading required package: Matrix  
##  
## Attaching package: 'Matrix'  
##  
## The following objects are masked from 'package:tidyr':  
##  
##     expand, pack, unpack  
##  
## Loading required package: survival  
##  
## Attaching package: 'survey'  
##  
## The following object is masked from 'package:graphics':  
##  
##     dotchart
```

```
source('data/nlsy97-educational-data/nlsy97-educational-data.R')
```

```
# Race key:  
# 1 Black  
# 2 Hispanic  
# 3 Mixed Race (Non-Hispanic)  
# 4 Non-Black
```

```
# Filter out non-responses
```

```
new_data <- new_data %>%  
  mutate(degree_label = case_when(  
    CV_HIGHEST_DEGREE_EVER_EDT_2017 == 0 ~ "None",  
    CV_HIGHEST_DEGREE_EVER_EDT_2017 == 1 ~ "GED",  
    CV_HIGHEST_DEGREE_EVER_EDT_2017 == 2 ~ "HS Diploma",
```

```

CV_HIGHEST_DEGREE_EVER_EDT_2017 == 3 ~ "AA",
CV_HIGHEST_DEGREE_EVER_EDT_2017 == 4 ~ "BA",
CV_HIGHEST_DEGREE_EVER_EDT_2017 == 5 ~ "MA",
CV_HIGHEST_DEGREE_EVER_EDT_2017 == 6 ~ "PhD",
TRUE ~ NA_character_
))

# Descriptive characteristics of respondents

# Degree attained

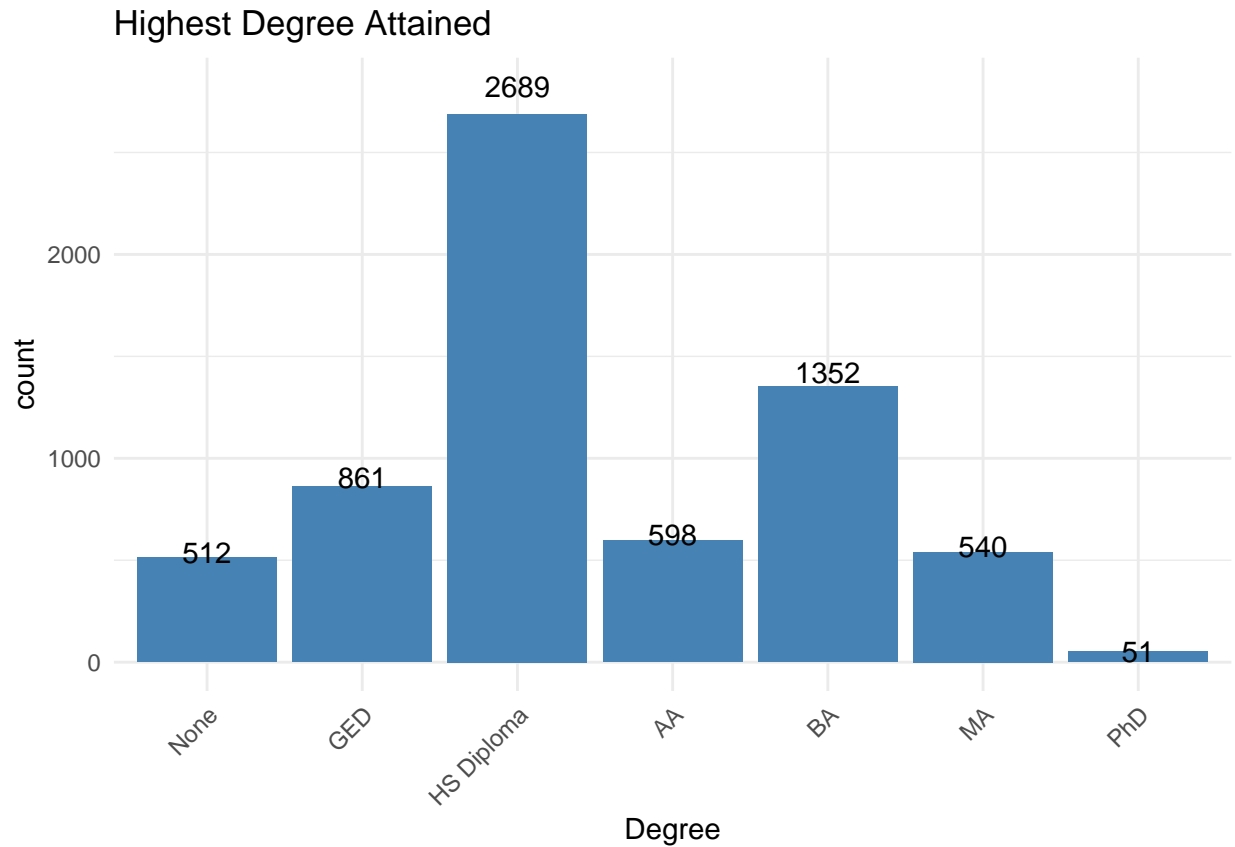
new_data <- new_data %>%
  mutate(degree_label = factor(degree_label,
                                levels = c("None", "GED", "HS Diploma",
                                             "AA", "BA", "MA", "PhD")))

new_data_rmNA <- new_data %>% dplyr::filter(!is.na(degree_label))

# Removes outliers
new_data_rmNA <- new_data_rmNA %>%
  filter(CV_HGC_RES_MOM_1997 <= 20 | is.na(CV_HGC_RES_MOM_1997)) %>%
  filter(CV_HGC_RES_DAD_1997 <= 20 | is.na(CV_HGC_RES_DAD_1997))

ggplot(new_data_rmNA, aes(x = degree_label)) +
  geom_bar(fill = "steelblue") +
  labs(title = "Highest Degree Attained",
       x = "Degree") +
  stat_count(geom = 'text',
             color = 'black',
             aes(label = after_stat(count)),
             position = position_stack(vjust = 1.05)) +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))

```

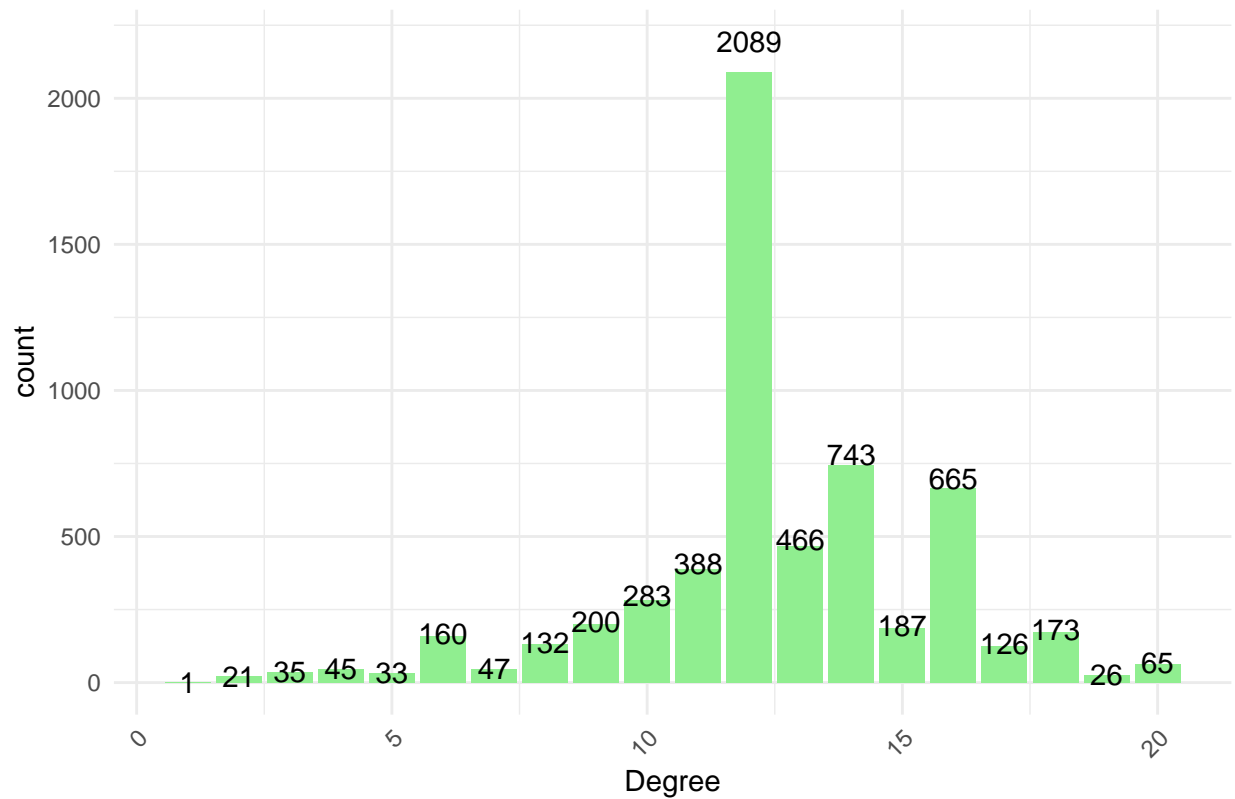


```
ggplot(new_data_rmNA, aes(x = CV_HGC_RES_MOM_1997)) +  
  geom_bar(fill = "lightgreen") +  
  labs(title = "Highest Degree Attained (Mother)",  
        x = "Degree") +  
  stat_count(geom = 'text',  
             color = 'black',  
             aes(label = after_stat(count)),  
             position = position_stack(vjust = 1.05)) +  
  theme_minimal() +  
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```

```
## Warning: Removed 718 rows containing non-finite outside the scale range  
## ('stat_count()').
```

```
## Warning: Removed 718 rows containing non-finite outside the scale range  
## ('stat_count()').
```

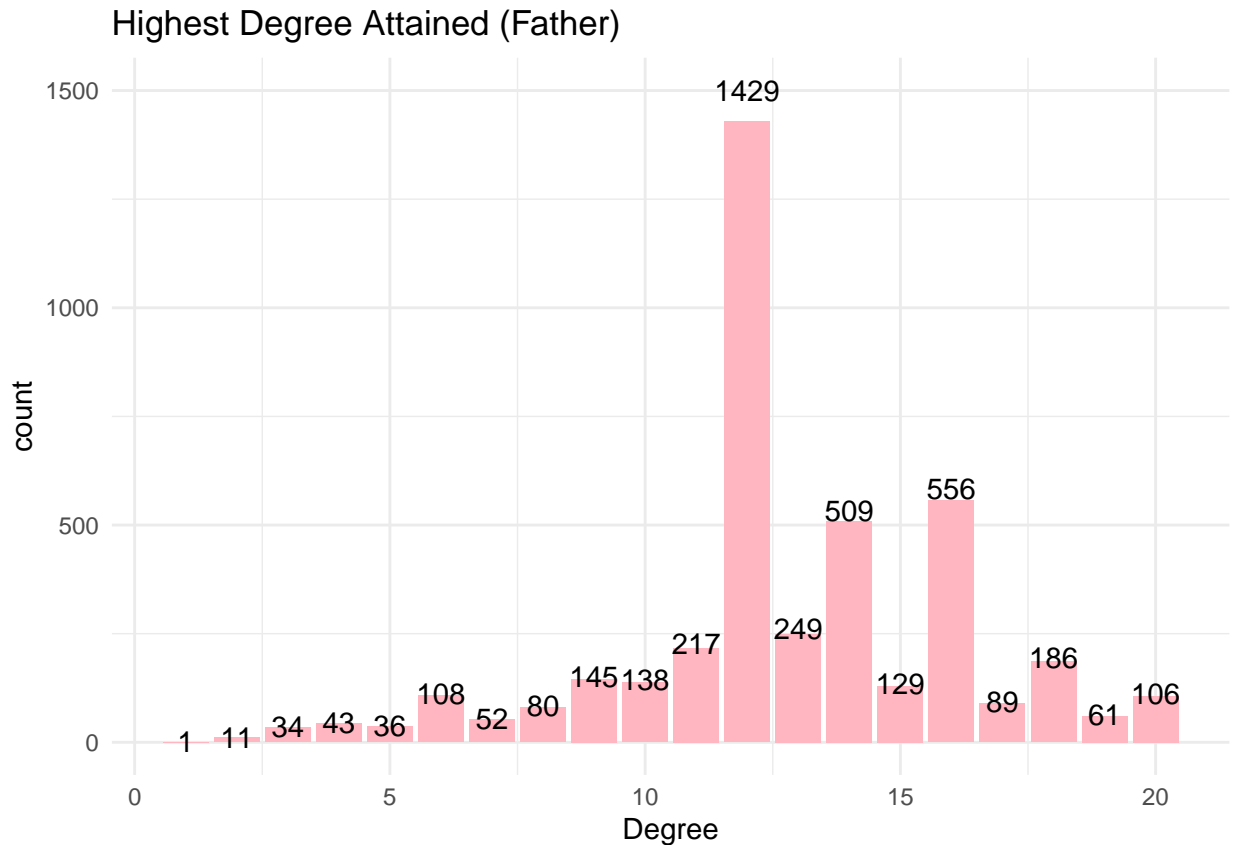
Highest Degree Attained (Mother)



```
ggplot(new_data_rmNA, aes(x = CV_HGC_RES_DAD_1997)) +
  geom_bar(fill = "lightpink") +
  labs(title = "Highest Degree Attained (Father)",
       x = "Degree") +
  stat_count(geom = 'text',
            color = 'black',
            aes(label = after_stat(count)),
            position = position_stack(vjust = 1.05)) +
  theme_minimal() +
  theme()
```

```
## Warning: Removed 2424 rows containing non-finite outside the scale range
## ('stat_count()').
```

```
## Warning: Removed 2424 rows containing non-finite outside the scale range
## ('stat_count()').
```

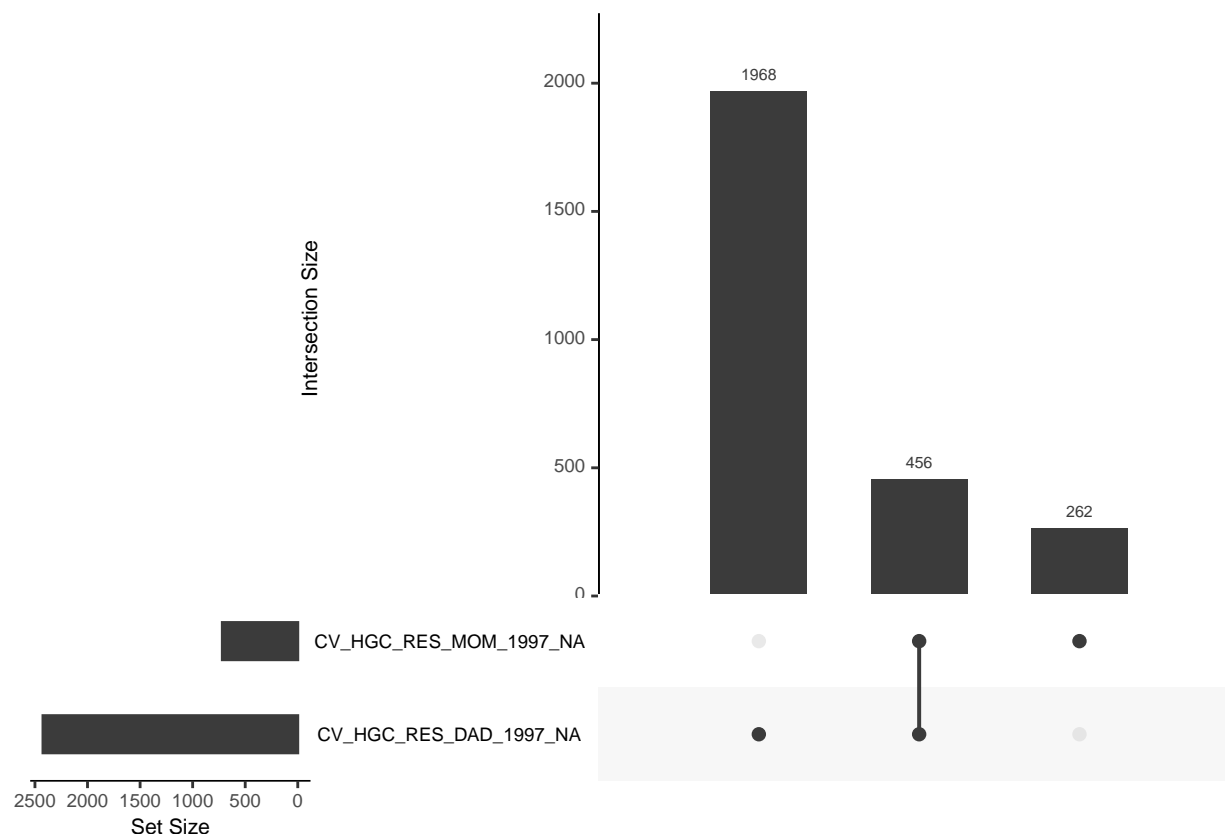


```
##### Start of missing data analysis #####
# Looking at missing-ness by race/ethnicity
new_data_rmNA %>%
  dplyr::select(KEY_RACE_ETHNICITY_1997, CV_HGC_RES_MOM_1997, CV_HGC_RES_DAD_1997) %>%
  mutate(
    mom_missing = is.na(CV_HGC_RES_MOM_1997),
    dad_missing = is.na(CV_HGC_RES_DAD_1997)
  ) %>%
  group_by(KEY_RACE_ETHNICITY_1997) %>%
  summarise(
    n = n(),
    mom_missing_pct = mean(mom_missing) * 100,
    dad_missing_pct = mean(dad_missing) * 100
  )
```

```
## # A tibble: 4 x 4
##   KEY_RACE_ETHNICITY_1997      n mom_missing_pct dad_missing_pct
##             <int> <int>          <dbl>          <dbl>
## 1                 1  1808             14.7             59.9
## 2                 2  1391             12.4             40.2
## 3                 3    62             27.4             43.5
## 4                 4  3342              7.90             22.6
```

```
# Testing for patterns in missing-ness

gg_miss_upset(new_data_rmNA)
```



```
mcar_data <- new_data_rmNA %>%
  dplyr::select(CV_HIGHEST_DEGREE_EVER_EDT_2017,
    CV_HGC_RES_MOM_1997,
    CV_HGC_RES_DAD_1997)

mcar_test(mcar_data)
```

```
## # A tibble: 1 x 4
##   statistic    df p.value missing.patterns
##   <dbl> <dbl>   <dbl>         <int>
## 1    364.     5     0.000         4
```

```
##### Imputations #####
# Multiple imputations
```

```
imp_data <- new_data_rmNA %>%
  dplyr::select(CV_HIGHEST_DEGREE_EVER_EDT_2017,
    KEY_RACE_ETHNICITY_1997, CV_HGC_RES_MOM_1997,
    CV_HGC_RES_DAD_1997, SAMPLING_WEIGHT_CC_2017)

imp <- mice(imp_data, m = 5, method = 'pmm')
```

```
##
## iter imp variable
## 1 1 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
```



```
## 1 2 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 1 3 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 1 4 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 1 5 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 2 1 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 2 2 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 2 3 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 2 4 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 2 5 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 3 1 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 3 2 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 3 3 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 3 4 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 3 5 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 4 1 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 4 2 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 4 3 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 4 4 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 4 5 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 5 1 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 5 2 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 5 3 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 5 4 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
## 5 5 CV_HGC_RES_MOM_1997 CV_HGC_RES_DAD_1997
```

```
imp <- complete(imp, action = "long", include = TRUE)

imp$CV_HIGHEST_DEGREE_EVER_EDT_2017 <- factor(
  imp$CV_HIGHEST_DEGREE_EVER_EDT_2017,
  levels = 0:6,
  labels = c("None", "GED", "HS", "AA", "BA", "MA", "PhD"),
  ordered = TRUE
)

# Re-convert to mids object
imp <- as.mids(imp)

# Runs ordinal log regr on imp data
pom_imp <- with(imp, polr(
  CV_HIGHEST_DEGREE_EVER_EDT_2017 ~ CV_HGC_RES_MOM_1997 +
    KEY_RACE_ETHNICITY_1997 + CV_HGC_RES_DAD_1997,
  Hess = TRUE
))

# Pool the results
pom_pooled <- pool(pom_imp)
summary(pom_pooled)
```

##		term	estimate	std.error	statistic	df
## 1		CV_HGC_RES_MOM_1997	0.1124828	0.01316745	8.542487	26.12942
## 2		KEY_RACE_ETHNICITY_1997	0.1573360	0.01925781	8.169982	198.95235
## 3		CV_HGC_RES_DAD_1997	0.1464721	0.01376937	10.637533	12.98876
## 4		None GED	0.9374694	0.11677139	8.028245	713.19784
## 5		GED HS	2.1534857	0.11516447	18.699219	515.45461

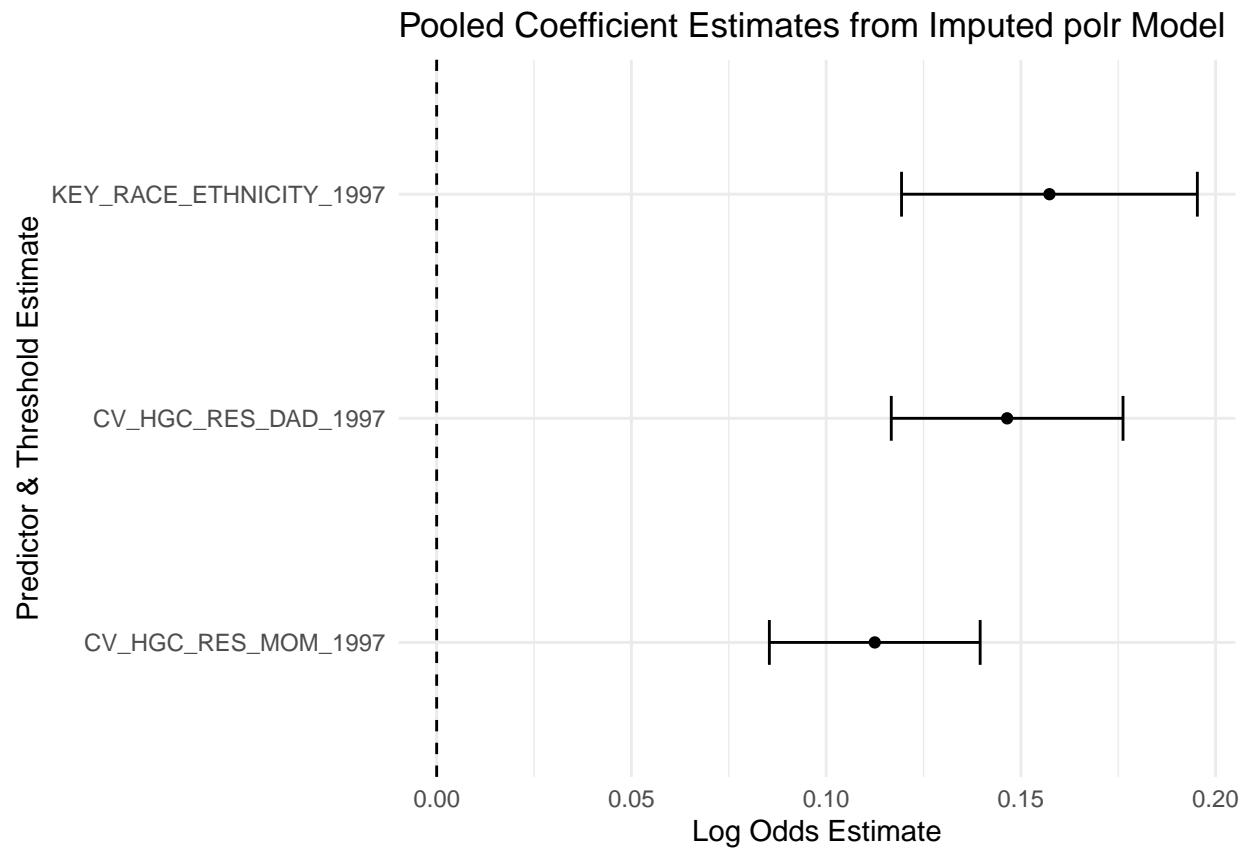
```
## 6          HS|AA 4.1927522 0.12375241 33.880165 513.50049
## 7          AA|BA 4.6604037 0.12656825 36.821271 492.42738
## 8          BA|MA 6.2598687 0.13769654 45.461336 478.72429
## 9          MA|PhD 8.8617108 0.19426315 45.617045 1371.14852
##          p.value
## 1 4.857036e-09
## 2 3.525224e-14
## 3 8.814373e-08
## 4 4.078123e-15
## 5 6.060005e-60
## 6 5.048183e-133
## 7 1.562305e-143
## 8 8.059260e-176
## 9 3.401583e-277
```

```
# Convert pooled polr results to tidy format
pooled_summary <- summary(pom_pooled)

# Add term names
tidy_pooled <- tidy(pom_pooled, conf.int = TRUE, conf.level = 0.95)

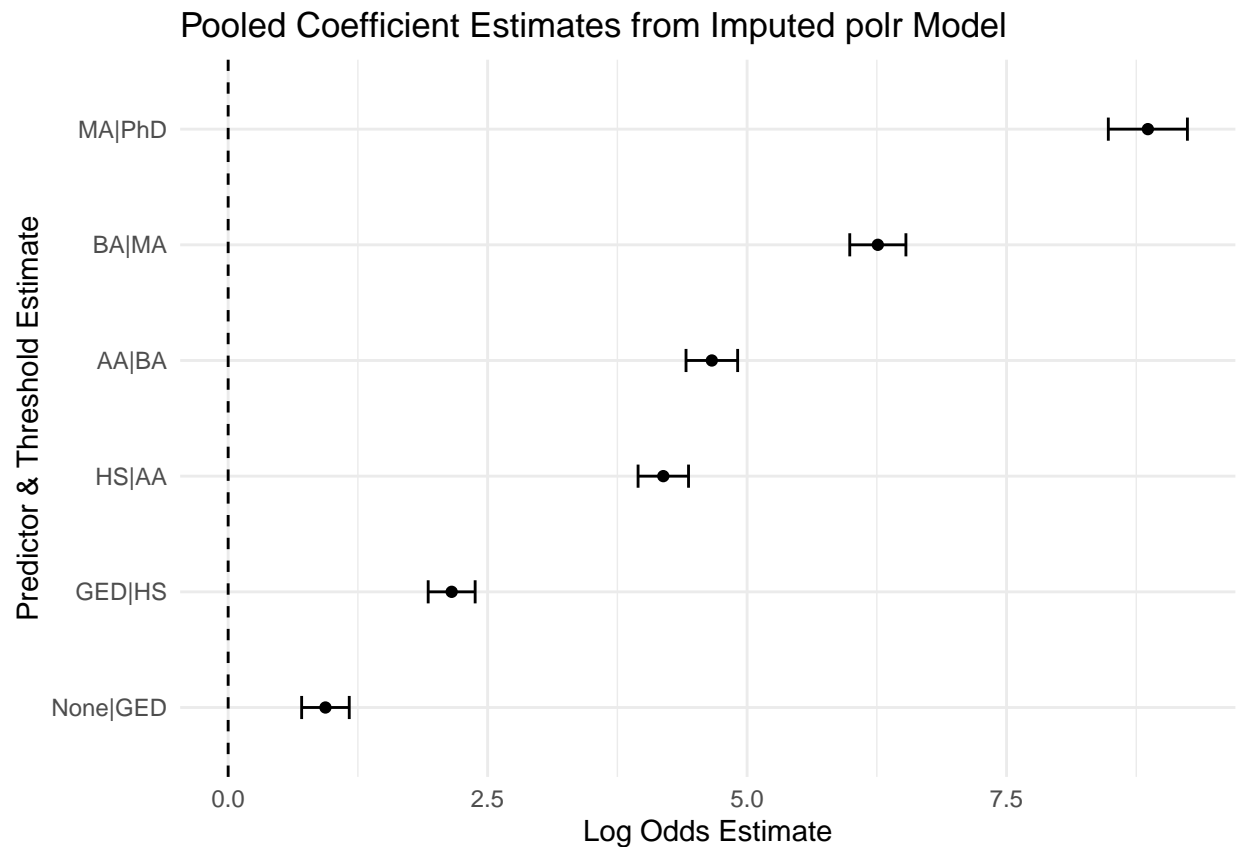
tidy_pooled_sub <- subset(tidy_pooled, tidy_pooled$estimate <= 0.8)

# Plot
ggplot(tidy_pooled_sub, aes(x = estimate, y = reorder(term, estimate))) +
  geom_point() +
  geom_errorbarh(aes(xmin = conf.low, xmax = conf.high), height = 0.2) +
  geom_vline(xintercept = 0, linetype = "dashed") +
  labs(
    title = "Pooled Coefficient Estimates from Imputed polr Model",
    x = "Log Odds Estimate",
    y = "Predictor & Threshold Estimate"
  ) +
  theme_minimal()
```



```
tidy_pooled_sub <- subset(tidy_pooled, tidy_pooled$estimate > 0.8)

# Plot
ggplot(tidy_pooled_sub, aes(x = estimate, y = reorder(term, estimate))) +
  geom_point() +
  geom_errorbarh(aes(xmin = conf.low, xmax = conf.high), height = 0.2) +
  geom_vline(xintercept = 0, linetype = "dashed") +
  labs(
    title = "Pooled Coefficient Estimates from Imputed polr Model",
    x = "Log Odds Estimate",
    y = "Predictor & Threshold Estimate"
  ) +
  theme_minimal()
```



```
# Selects an imp
completed_data <- complete(imp, action = 5L)

# Weighted GLM
completed_data$degree_num <- as.numeric(
  completed_data$CV_HIGHEST_DEGREE_EVER_EDT_2017)

svy_design <- svydesign(
  ids = ~1,
  weights = ~SAMPLING_WEIGHT_CC_2017,
  data = completed_data
)

svy_model <- svyglm(
  degree_num ~ CV_HGC_RES_MOM_1997 +
    KEY_RACE_ETHNICITY_1997 + CV_HGC_RES_DAD_1997,
  design = svy_design
)

summary(svy_model)

##
## Call:
## svyglm(formula = degree_num ~ CV_HGC_RES_MOM_1997 + KEY_RACE_ETHNICITY_1997 +
##   CV_HGC_RES_DAD_1997, design = svy_design)
##
```

```
## Survey design:
## svydesign(ids = ~1, weights = ~SAMPLING_WEIGHT_CC_2017, data = completed_data)
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.600724   0.084230   7.132 1.10e-12 ***
## CV_HGC_RES_MOM_1997 0.099820   0.008536  11.694 < 2e-16 ***
## KEY_RACE_ETHNICITY_1997 0.089688   0.013247   6.770 1.39e-11 ***
## CV_HGC_RES_DAD_1997  0.110427   0.007290  15.148 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 1.635078)
##
## Number of Fisher Scoring iterations: 2
```

```
# Visualize pooled regression coefficients of imps (not done yet)
```

```
library(effects)
```

```
## Warning: package 'effects' was built under R version 4.4.3
```

```
## Loading required package: carData
```

```
## Warning: package 'carData' was built under R version 4.4.3
```

```
## lattice theme set by effectsTheme()
```

```
## See ?effectsTheme for details.
```

```
# Use one completed data set as demonstration
```

```
effs <- Effect(c("KEY_RACE_ETHNICITY_1997", "CV_HGC_RES_MOM_1997"),
               svy_model)
```

```
# For predicted probabilities
```

```
as.data.frame(effs)
```

	KEY_RACE_ETHNICITY_1997	CV_HGC_RES_MOM_1997	fit	se	lower
## 1	1.0	1.0	2.229896	0.10281198	2.028351
## 2	1.8	1.0	2.301646	0.10217012	2.101360
## 3	2.5	1.0	2.364428	0.10250789	2.163479
## 4	3.2	1.0	2.427209	0.10367725	2.223968
## 5	4.0	1.0	2.498960	0.10599574	2.291174
## 6	1.0	5.8	2.709034	0.06464845	2.582302
## 7	1.8	5.8	2.780784	0.06272719	2.657818
## 8	2.5	5.8	2.843566	0.06248859	2.721068
## 9	3.2	5.8	2.906347	0.06361540	2.781640
## 10	4.0	5.8	2.978098	0.06648226	2.847771
## 11	1.0	10.0	3.128279	0.03663149	3.056470
## 12	1.8	10.0	3.200030	0.03159326	3.138097
## 13	2.5	10.0	3.262811	0.02969215	3.204605
## 14	3.2	10.0	3.325593	0.03061186	3.265584
## 15	4.0	10.0	3.397343	0.03480013	3.329124

## 16	1.0	15.0 3.627381 0.03627416 3.556272
## 17	1.8	15.0 3.699132 0.02922695 3.641837
## 18	2.5	15.0 3.761913 0.02519067 3.712531
## 19	3.2	15.0 3.824695 0.02422596 3.777204
## 20	4.0	15.0 3.896445 0.02725787 3.843011
## 21	1.0	20.0 4.126483 0.07023698 3.988796
## 22	1.8	20.0 4.198233 0.06598307 4.068885
## 23	2.5	20.0 4.261015 0.06349005 4.136554
## 24	3.2	20.0 4.323797 0.06229108 4.201686
## 25	4.0	20.0 4.395547 0.06259716 4.272836
##	upper	
## 1	2.431441	
## 2	2.501933	
## 3	2.565376	
## 4	2.630450	
## 5	2.706746	
## 6	2.835766	
## 7	2.903750	
## 8	2.966063	
## 9	3.031054	
## 10	3.108424	
## 11	3.200089	
## 12	3.261963	
## 13	3.321017	
## 14	3.385602	
## 15	3.465563	
## 16	3.698490	
## 17	3.756426	
## 18	3.811295	
## 19	3.872185	
## 20	3.949879	
## 21	4.264170	
## 22	4.327582	
## 23	4.385476	
## 24	4.445907	
## 25	4.518258	