

# Age and Education Differences Among Couples

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```
acs2021_couples$educ_numeric <- fct_recode(acs2021_couples$EDUC,
                                           "0" = "N/A or no schooling",
                                           "2" = "Nursery school to grade 4",
                                           "6.5" = "Grade 5, 6, 7, or 8",
                                           "9" = "Grade 9",
                                           "10" = "Grade 10",
                                           "11" = "Grade 11",
                                           "12" = "Grade 12",
                                           "13" = "1 year of college",
                                           "14" = "2 years of college",
                                           "15" = "3 years of college",
                                           "16" = "4 years of college",
                                           "17" = "5+ years of college")

acs2021_couples$educ_numeric <-
as.numeric(levels(acs2021_couples$educ_numeric))[acs2021_couples$educ_numeric]

acs2021_couples$h_educ_numeric <- fct_recode(acs2021_couples$h_educ,
                                              "0" = "N/A or no schooling",
                                              "2" = "Nursery school to grade 4",
                                              "6.5" = "Grade 5, 6, 7, or 8",
                                              "9" = "Grade 9",
                                              "10" = "Grade 10",
                                              "11" = "Grade 11",
                                              "12" = "Grade 12",
                                              "13" = "1 year of college",
                                              "14" = "2 years of college",
                                              "15" = "3 years of college",
                                              "16" = "4 years of college",
                                              "17" = "5+ years of college")

acs2021_couples$h_educ_numeric <-
as.numeric(levels(acs2021_couples$h_educ_numeric))[acs2021_couples$h_educ_numeric]

acs2021_couples$educ_diff <- acs2021_couples$educ_numeric -
acs2021_couples$h_educ_numeric

# Hypothesis:

#  $H_0: \beta_2 = \beta_3 = 0$  (the squared and cubic terms of education difference have
```

```

no joint effect)

# HA: At least one of  $\theta_2$ ,  $\theta_3 \neq 0$  (nonlinear relationship between age and
education differences)

# Model 1: Simple linear relationship between age and education differences

m1 <- lm(age_diff ~ educ_diff, data = acs2021_couples)

# Model 2: Add nonlinear terms for education difference

m2 <- lm(age_diff ~ educ_diff + I(educ_diff^2) + I(educ_diff^3),
data = acs2021_couples)

# Model 3: Add partner controls (ages and education levels)

acs2021_couples$avg_age <- (acs2021_couples$AGE + acs2021_couples$h_age) / 2

m3 <- lm(age_diff ~ educ_diff + I(educ_diff^2) + I(educ_diff^3) +
        avg_age + I(avg_age^2),
        data = acs2021_couples)

# Robust coefficient test for Model 1

coeftest(m1, vcov = vcovHC)

##
## t test of coefficients:
##
##              Estimate Std. Error   t value Pr(>|t|)
## (Intercept) -0.7556547  0.0072612 -104.0678  <2e-16 ***
## educ_diff    0.0072887  0.0041757   1.7455   0.0809 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# Wald test for nonlinearity (do polynomial terms jointly matter?)

waldtest(m1, m2, vcov = vcovHC)

## Wald test
##
## Model 1: age_diff ~ educ_diff
## Model 2: age_diff ~ educ_diff + I(educ_diff^2) + I(educ_diff^3)
##   Res.Df Df    F    Pr(>F)
## 1 776062
## 2 776060  2 368.8 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

*# Wald test for additional age/education controls*

```
waldtest(m2, m3, vcov = vcovHC)
```

```
## Wald test
```

```
##
```

```
## Model 1: age_diff ~ educ_diff + I(educ_diff^2) + I(educ_diff^3)
```

```
## Model 2: age_diff ~ educ_diff + I(educ_diff^2) + I(educ_diff^3) + avg_age  
+  
## I(avg_age^2)
```

```
## Res.Df Df F Pr(>F)
```

```
## 1 776060
```

```
## 2 776058 2 151.16 < 2.2e-16 ***
```

```
## ---
```

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

```
models <- list(  
  "M 1" = lm(age_diff ~ educ_diff, data = acs2021_couples),  
  "M 2" = lm(age_diff ~ educ_diff + I(educ_diff^2) + I(educ_diff^3),  
    data = acs2021_couples),  
  "M 3" = lm(age_diff ~ educ_diff + I(educ_diff^2) + I(educ_diff^3) +  
    avg_age + I(avg_age^2),  
    data = acs2021_couples)  
)
```

```
modelsummary(models, stars = TRUE, output = "default")
```

	M 1	M 2	M 3
(Intercept)	-0.756*** (0.007)	-0.650*** (0.008)	-0.087 (0.070)
educ_diff	0.007** (0.003)	-0.038*** (0.004)	-0.037*** (0.004)
I(educ_diff^2)		-0.014*** (0.000)	-0.014*** (0.000)
I(educ_diff^3)		0.000*** (0.000)	0.000*** (0.000)
avg_age			-0.030*** (0.003)
I(avg_age^2)			0.000*** (0.000)
Num.Obs.	776064	776064	776064
R2	0.000	0.003	0.003
R2 Adj.	0.000	0.003	0.003

	M 1	M 2	M 3
AIC	5085935.4	5083663.0	5083402.2
BIC	5085970.1	5083720.8	5083483.1
Log.Lik.	-2542964.709	-2541826.519	-2541694.079
F	7.212	762.314	510.528
RMSE	6.41	6.40	6.40

- $p < 0.1$ , \*  $p < 0.05$ , \*\*  $p < 0.01$ , \*\*\*  $p < 0.001$

*# Predicted values from Model 2*

```
plot_data <- acs2021_couples %>%
mutate(pred_age_diff = predict(m2))

ggplot(plot_data, aes(x = educ_diff, y = age_diff)) +
  geom_point(alpha = 0.1, color = "gray60") +
  geom_line(aes(y = pred_age_diff), color = "blue", linewidth = 1.1) +
  theme_minimal() +
  labs(
    title = "Relationship Between Education Difference and Age Difference",
    subtitle = "Predicted values from polynomial model (M2)",
    x = "Education Difference (years of schooling)",
    y = "Age Difference (years)"
  )
```

Relationship Between Education Difference and Age Di  
Predicted values from polynomial model (M2)

