Regression diagnostics

Statistical Reasoning
and Quantitative Methods

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Session 11

Outline

Diagnostics

Regression models produce **fitted** (predicted) values and residuals that hold the unexplained variance for each data point.
Issues that arise in that context are:

- **unreliable coefficients** due to multicollinearity, i.e. interactions between independent variables
- **unreliable significance tests** due to heteroskedasticity, i.e. heterogeneous variance in the residuals
- **unreliable predictions** due to outliers and influential points in the data that either do not fit or 'overfit' the model

Note: the model still assumes a **linear, additive** relationship between Y and $X_1, X_2, \ldots X_k$. That assumption can also be violated among other matters.

Fitting a **multiple** linear regression model

The model also fits a **linear function** to the data, of the form:

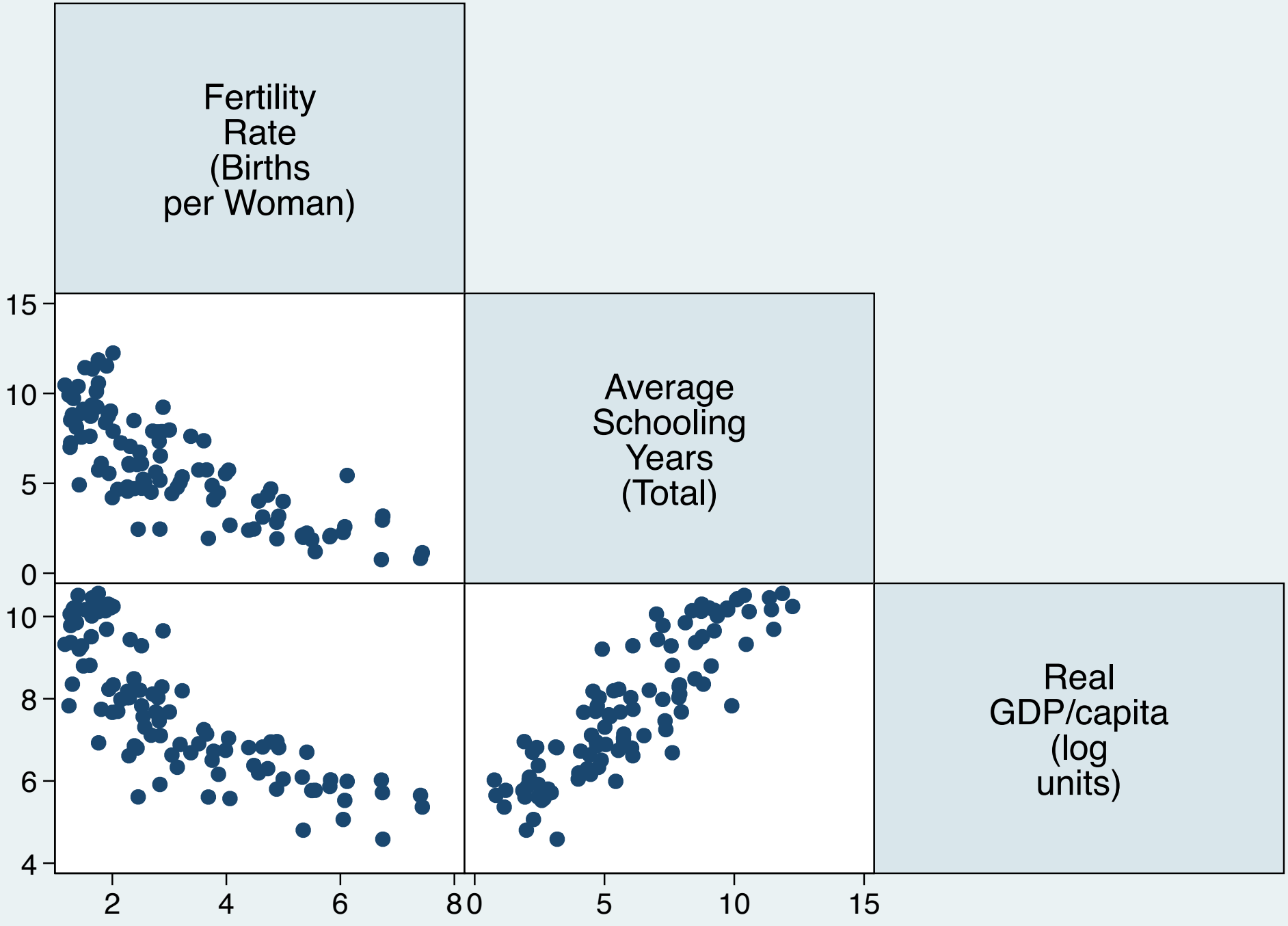
$$Y = \alpha + \beta_1 X_1 + \beta_2 X_2 + \ldots + \beta_k X_k + \epsilon$$

where:

- Y is the **dependent variable** (response)
- X is a **vector** of **independent variables** (predictors)
- α is the **constant**
- $\beta_1 X_1 + \beta_2 X_2 + \ldots + \beta_k X_k$ is a **vector** of **regression coefficients**
- ϵ is the **error term** (residuals)

Note: the model assumes that the relationship is **linear** and **additive**.

The estimation of regression coefficients in a k-dimensional space is computationally more intensive, but is also based on least squares.


Fertility Rate (Births per Woman)
Average Schooling Years (Total)
Real GDP/capita (log units)
15
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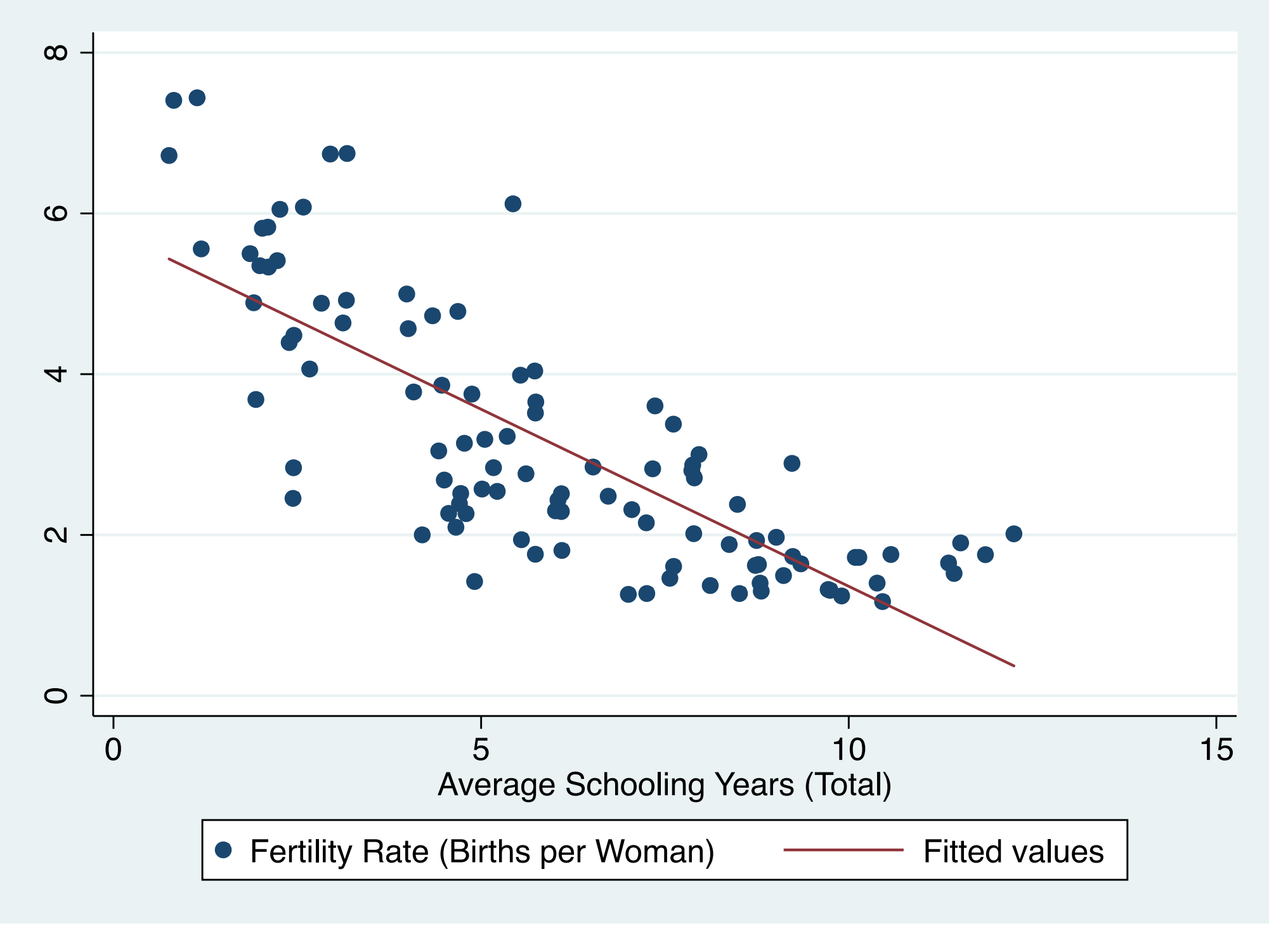
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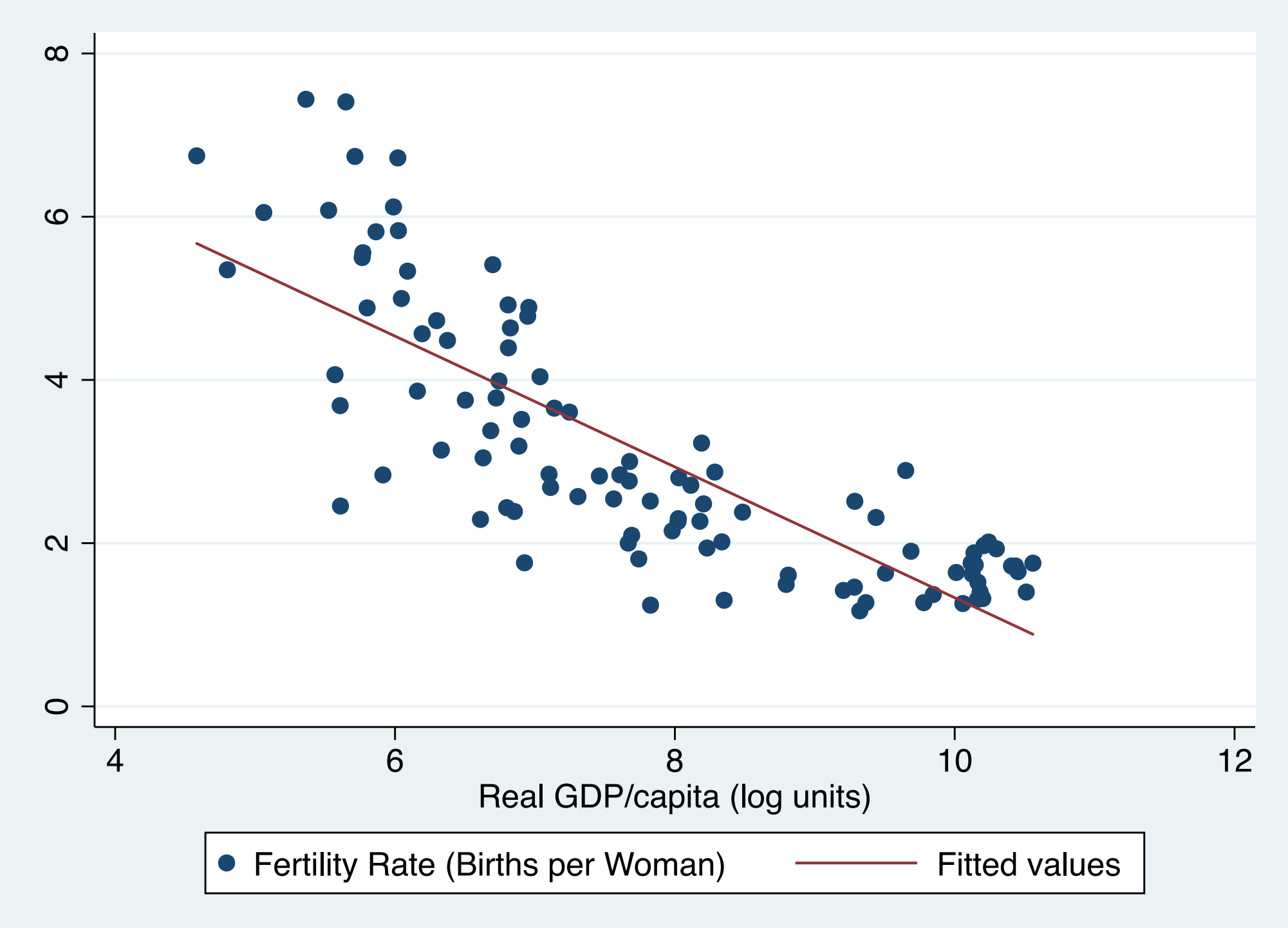
Average Schooling Years (Total)
Fertility Rate (Births per Woman)
Fitted values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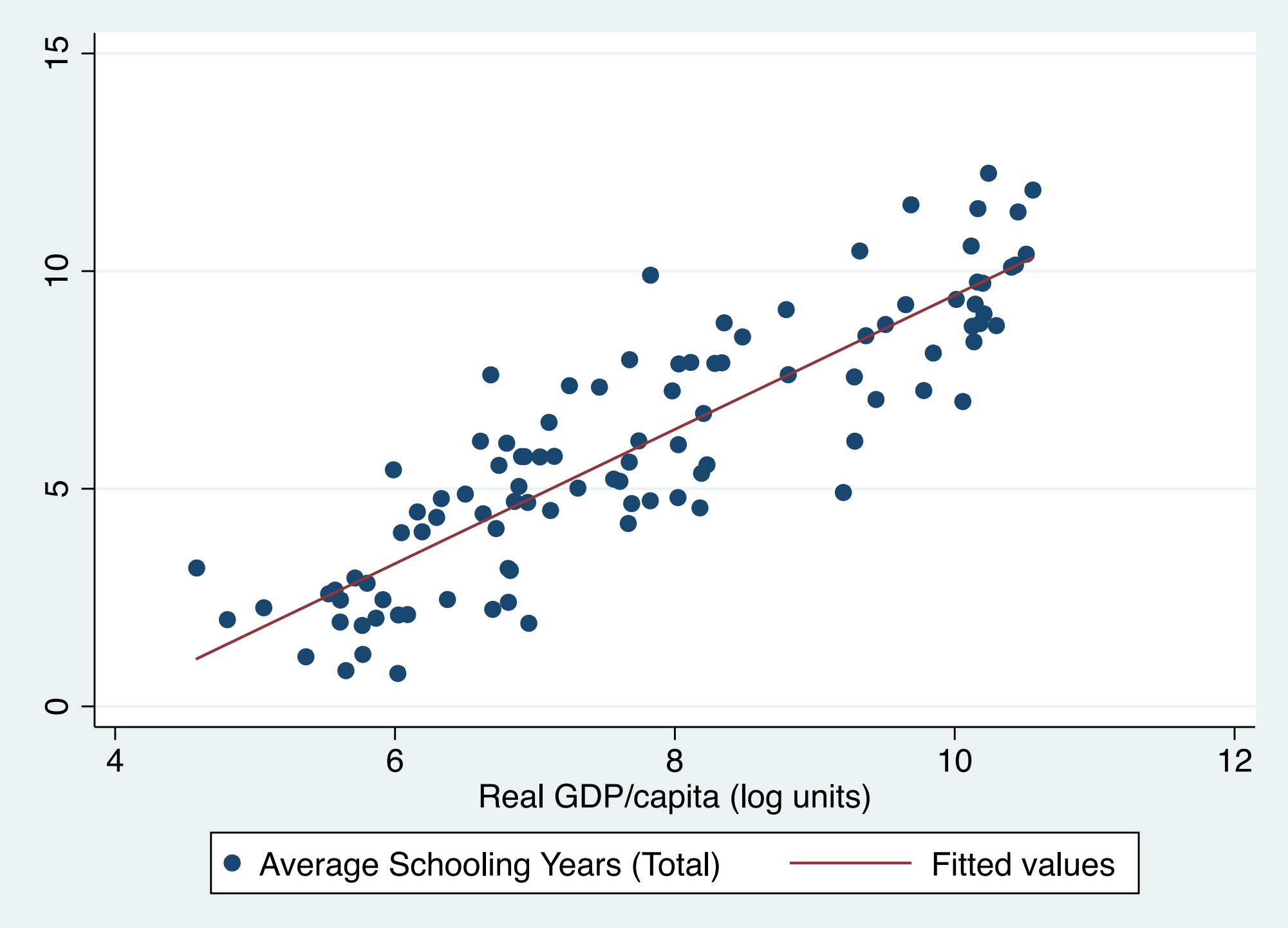
Real GDP/capita (log units)
Fertility Rate (Births per Woman)
Fitted values


15
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12
Real GDP/capita (log units)
Average Schooling Years (Total)
Fitted values

Multiple regression output

`reg births schooling log_gdpc`

The `reg` command can take any number of **continuous** variables as arguments, and shows unstandardised coefficients by default, using their original metric and possible transformation:

```
. reg births schooling log_gdpc
```

Source	SS	df	MS
Model	150.301883	2	75.1509417
Residual	70.475313	83	.849100157
Total	220.777196	85	2.59737878

Number of obs	86
F(2, 83)	88.51
Prob > F	0.0000
R-squared	0.6808
Adj R-squared	0.6731
Root MSE	.92147

births	Coef.	Std. Err.	t	P>|t|	[95% Conf.	Interval]
schooling	-.1976117	.0724595	-2.73	0.008	-.3417306	-.0534927
log_gdpc	-.4703416	.1324501	-3.55	0.001	-.7337796	-.2069036
_cons	7.950304	.6861182	11.59	0.000	6.585642	9.314965

Standardised coefficients

```
reg births schooling log_gdpc, beta
```

The `beta` option provides standardised coefficients, which use the **standard deviation of regressors** (or predictor, i.e. the independent variables) in order to provide coefficients with comparable units:

births	Coef.	Std. Err.	t	P>|t|	Beta
schooling	-.1976117	.0724595	-2.73	0.008	-.3686479
log_gdpc	-.4703416	.1324501	-3.55	0.001	-.4800156
_cons	7.950304	.6861182	11.59	0.000	.

(identical output for overall model fit omitted)

Dummies

```
reg births schooling i.region
```

Categorical variables can be used as dummies, i.e. binary recodes of each category that are tested against a **reference category** to provide regression coefficients for net effect of that category alone:

births	Coef.	Std. Err.	t	P>|t|	[95% Conf.	Interval]
schooling	-.0415563	.0639718	-0.65	0.518	-.1688888	.0857763
log_gdpc	-.742187	.1380037	-5.38	0.000	-1.016876	-.4674975
region						
2	-.6523485	.5803126	-1.12	0.264	-1.807432	.5027349
3	.3682404	.254364	1.45	0.152	-.1380585	.8745393
4	1.411177	.2486027	5.68	0.000	.9163457	1.906008
5	1.167491	.337383	3.46	0.001	.4959471	1.839035
_cons	8.315004	.8006456	10.39	0.000	6.721359	9.908649

(identical output for overall model fit omitted)