

ECON 0150 | Economic Data Analysis

The economist's data analysis pipeline.

Part 5.4 | Model Selection

Model Selection

How do we know if adding a variable improves our model?

- *Adding variables reduces SSE (better fit to the data)*
- *But is the improvement real or just fitting noise?*
- *We need a way to test whether the improvement is statistically significant*

R^2 : Proportion of Variation Explained

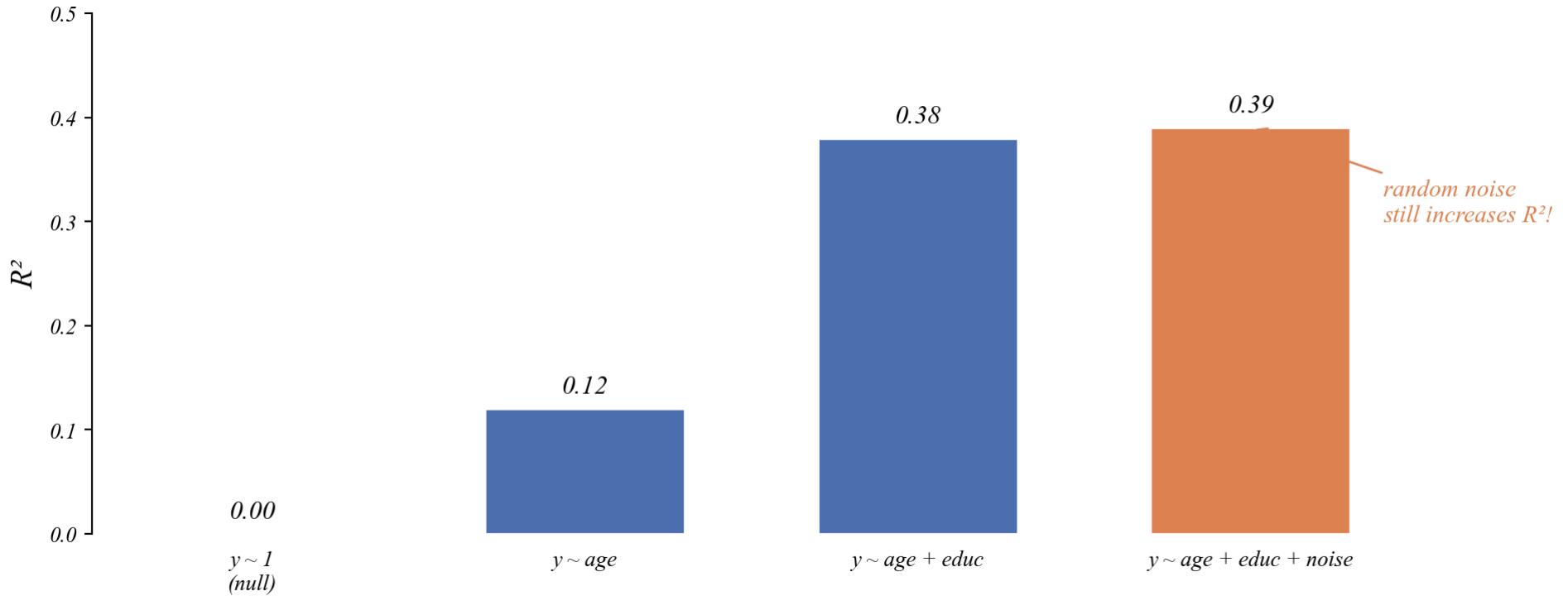
How much of the variation does our model capture?

$$R^2 = 1 - \frac{SSE}{SST}$$

- SST : total variation (SSE of the null model—just the mean)
- SSE : leftover variation after fitting the model
- $R^2 = 0$: model does no better than the mean
- $R^2 = 1$: model predicts perfectly

The Problem with R^2

R^2 always goes up when you add variables.



- > even adding random noise will reduce SSE a little
- > so how do we know if the improvement is real?

The F-Test

Is the reduction in SSE larger than we'd expect by chance?

Compare two models:

- **Restricted:** fewer variables → higher SSE_R
- **Full:** more variables → lower SSE_F

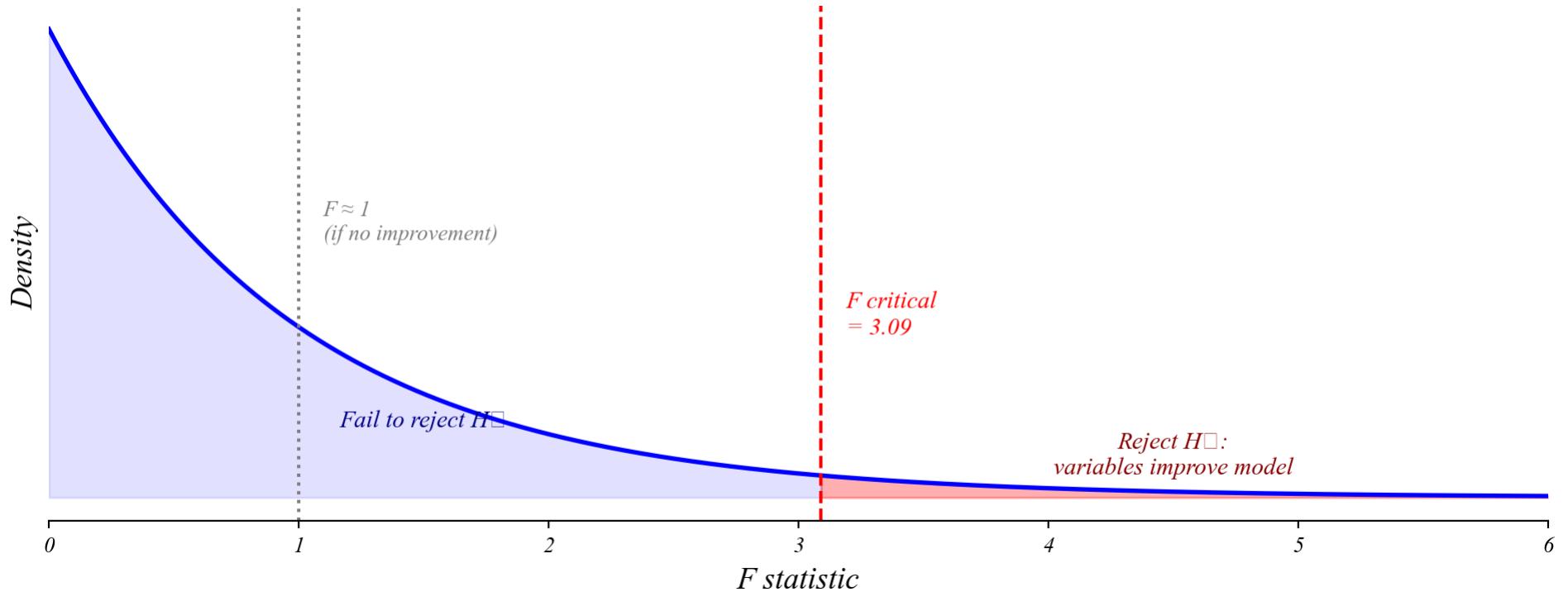
$$F = \frac{(SSE_R - SSE_F)/(df_R - df_F)}{SSE_F/df_F}$$

> numerator: average SSE reduction per variable added

> denominator: average remaining error

The F-Test

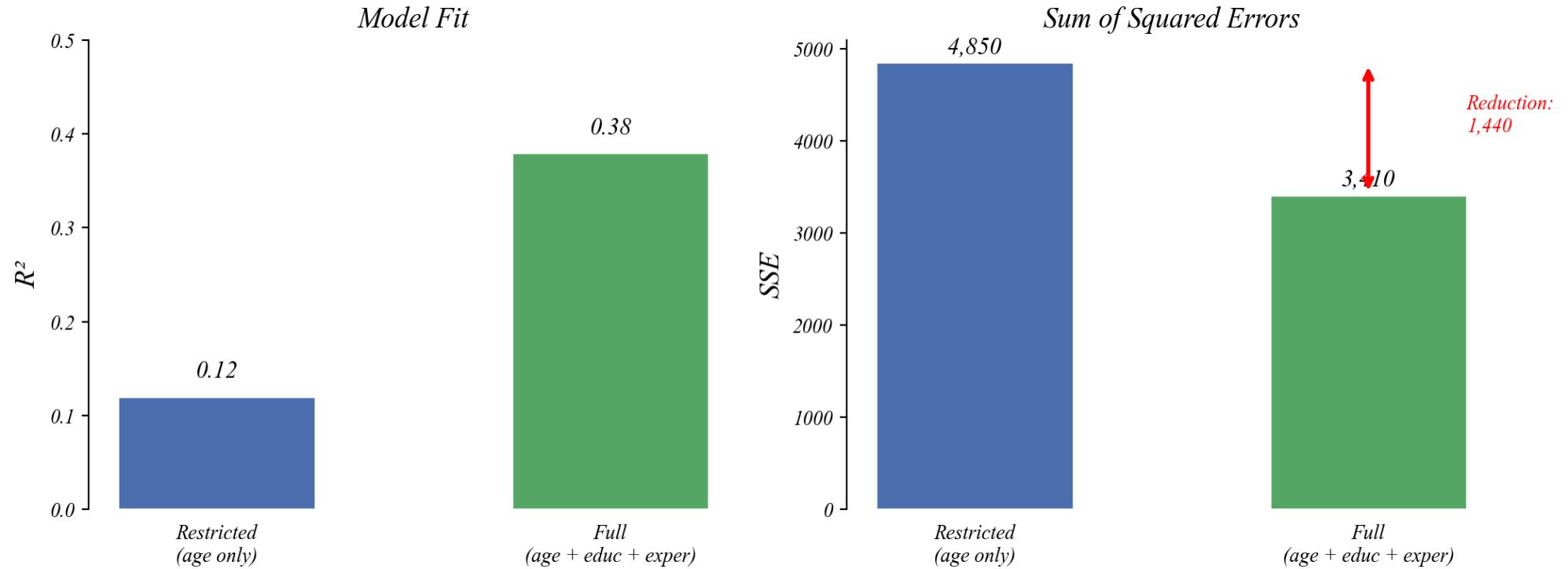
If the new variables are just noise, $F \approx 1$. If meaningful, F is large.



> large F means the SSE reduction is unlikely due to chance

Example: Wage Model

Does adding education and experience improve predictions?



$> R^2$ increased by 0.26—but is that significant?

Example: Wage Model

The F-test tells us if the improvement is statistically significant.

```
1 from scipy import stats
2
3 # Model comparison
4 sse_r = 4850      # restricted model (age only)
5 sse_f = 3410      # full model (age + educ + exper)
6 k = 2             # variables added
7 n = 100           # sample size
8
9 # F-statistic
10 f_stat = ((sse_r - sse_f) / k) / (sse_f / (n - 4))
11 p_value = 1 - stats.f.cdf(f_stat, k, n - 4)
```

F-statistic: 20.27

p-value: 0.0000

$> p < 0.05$: education and experience significantly improve the model

Summary

Model selection uses the same logic as hypothesis testing.

- R^2 tells us how much variation the model explains
- R^2 always increases when adding variables—even useless ones
- F-test asks: is the SSE reduction statistically significant?
- Same idea as t-tests, but for groups of variables