

Testing for Heteroskedasticity

Homoskedasticity is a strong assumption and is rarely realistic for data sets in this course.

- This isn't a problem for applied work because we have techniques that are robust to heteroskedasticity.
- Our formula for the variance of OLS coefficients is wrong under heteroskedasticity, but heteroskedasticity-robust standard errors can be used instead.
 - Other names: Huber-White standard errors, Eicker-White, or Eicker-Huber-White
 - These researchers developed an expression for variance estimates that are unbiased without homoskedasticity.

Testing for Heteroskedasticity (cont.)

- We can work around heteroskedasticity, but it's worth discussing how to test for it.
- Usually the best thing to do is to look at the diagnostic plots.
- R provides two plots that can help identify heteroskedasticity.

Note: If there are more data points on one side of the plot, it may make it look like there's more variation than is actually occurring.

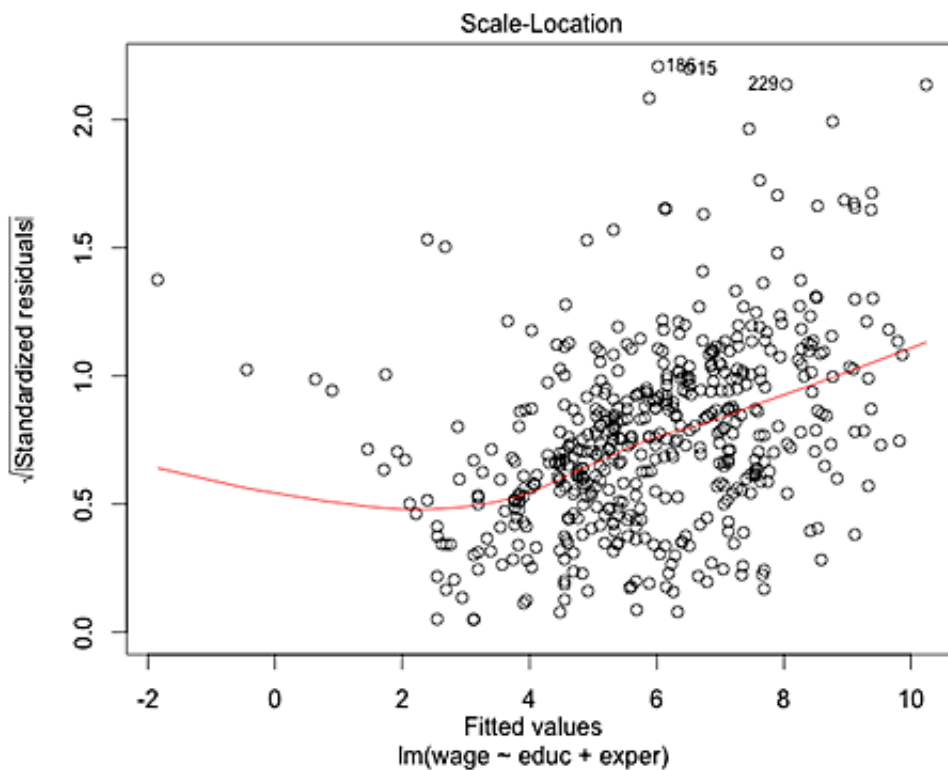
Testing Homoskedasticity: Residuals vs. Fitted-Values Plot

- Example: Fitted wage model, using data from 1976.
 - $\text{wage} = -3.39 + 0.644\text{educ} + 0.070\text{exper} + u$
- If errors were homoskedastic, the band would be of uniform thickness.
- Remember, the red line is a smoother that approximates the mean of the residuals.
 - It appears to show a violation of zero-conditional mean.
 - But the spline does not help with heteroskedasticity because that is about the deviations from the mean.

Scale-Location Plot

R provides something called a scale-location plot.

- This is related to the residual vs. fitted-value plot, but the residuals are transformed in two ways:
 - First, we calculate the absolute value.
 - More variance appears as points higher on the y -axis.
 - At this point, there is likely to be a lot of skew, making it tough to inspect visually.
 - Next, we calculate the square root.
 - This reduces skew, moving points away from the x -axis.
 - After these transformations, homoskedastic errors should appear as a horizontal band of points.



Breusch-Pagan Test

- The Breusch-Pagan test is a statistical test that can be used to check for heteroskedasticity.
- The null hypothesis is that there is homoskedasticity.
- If there is a significant result, that is evidence supporting heteroskedasticity.
- Warning: As with most tests of this nature, sample size matters greatly.
 - For a few hundred data points or more, almost any amount of heteroskedasticity will appear as a significant test.
 - For small data sets, the test will rarely be significant.

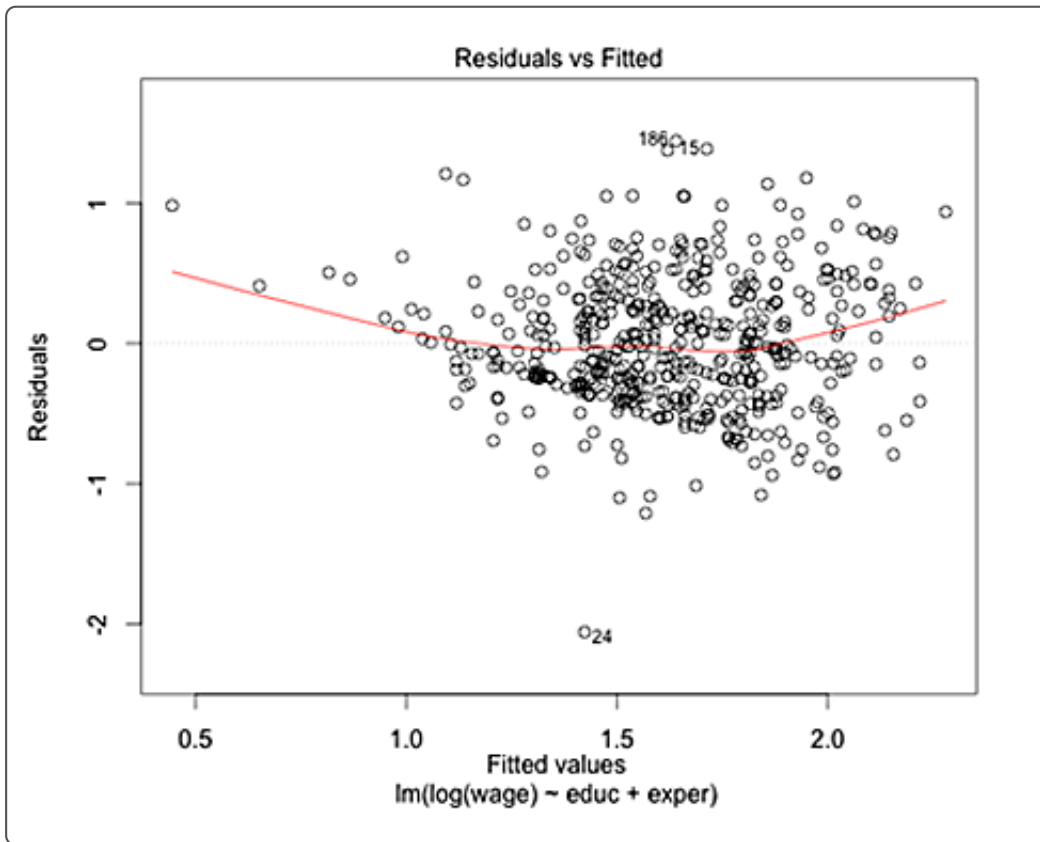
With the Breusch-Pagan test, it's best to keep sample size in mind and to use it in conjunction with diagnostic plots.

Responding to Heteroskedasticity

- The simplest solution is to switch to heteroskedasticity-robust tools.
- In particular, White standard errors are robust to heteroskedasticity.
 - Many researchers recommend using these by default.
 - If there is homoskedasticity, White standard errors will be larger (more conservative) than regular standard errors, so it is slightly in your favor to use regular errors.

Responding to Heteroskedasticity (cont.)

- At times, heteroskedasticity accompanies a violation of zero-conditional mean.
 - There may be an exponential relationship in the data, especially if $\text{var}(y)$ seems correlated with $E(y)$.
- This seemed to happen for the previous wage regression.
 - There are a lot of outliers in the positive direction, and there is also more variance on this side.
- More commonly, researchers will model the log of wage.



Plots With Natural Log Wage

- Using the same model as previously, replacing wage with the natural log of wage.
 - $\text{Log}(\hat{\text{wage}}) = 0.21 + 0.09\text{educ} + 0.010\text{exper}$
 - Notice that the residual vs. fitted-value plot looks more even, and the scale-location plot appears flatter.

Log(wage) and Intuition

- Warning: In this case, it makes intuitive sense to take the $\log(\text{wage})$.
 - Everyone has a sense for what a 10-percent increase in wage means.
 - A \$100 increase could be very different for different people.
- Do not take the log of a variable just to fix heteroskedasticity if it doesn't make theoretical sense.
 - We want our specification to be guided by theory and the tests we wish to perform, not by what choices result in homoskedasticity.
 - That's why the best practice is simply to use heteroskedasticity-robust standard errors.