

Partial (sequential) F tests

$$Y = \beta_0 + \beta_1 X_1 + \varepsilon \quad (1)$$

$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \varepsilon \quad (2)$$

A sequential F test is a partial F test for the addition of a single term to an existing model

→ $+ \beta_2 X_2$

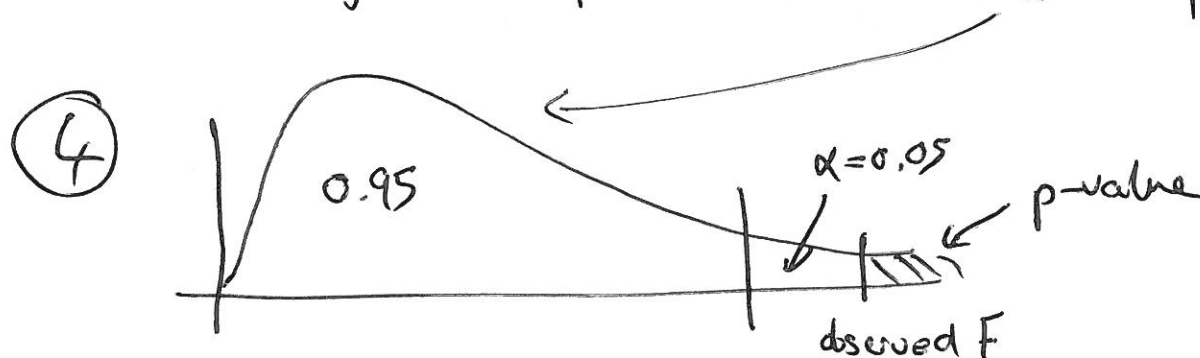
F test

① $H_0: \frac{\sigma^2_{\text{addition}}}{\sigma^2_{\text{Error, larger model}}} = 1$ vs $H_A: \frac{\sigma^2_a}{\sigma^2_\varepsilon} > 1$
(in variance terms)

equiv. $H_0: \beta_2 = 0$ vs $H_A: \beta_2 \neq 0$
(in mean terms)

② Test statistic $F = \frac{MS_{\text{addition}}}{MS_{\text{error/residual}}} \sim F_{\text{addition df, error df}}$
 ← from model ① →

③ Decision rule $\alpha = 0.05$
 reject H_0 if observed $F > F_{\text{addition df, error df}}$



⑤ Compare p-value with α & draw a conclusion

Partial (nested) F test for a group of terms in a nested model

Model (A) $Y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + \varepsilon$ "base" model

Model (B) $Y = \underbrace{\hspace{10em}}_{\text{"base" model}} + \beta_4 x_4 + \beta_5 x_5 + \varepsilon$
 "expanded" model additions

"base" model (A) is nested inside the "expanded" model (B)
 model (A) is a subset of - model (B)

① $H_0: \frac{\sigma^2_{\text{addition}}}{\sigma^2_{\text{error, larger model (B)}}} = 1$

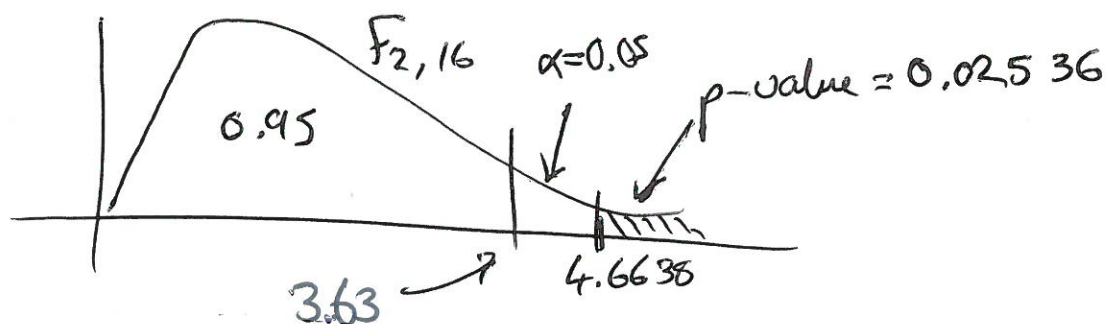
vs $H_A: \frac{\sigma^2_{\text{addition}}}{\sigma^2_{\text{error}}} > 1$

or equivalently $H_0: \beta_4 = \beta_5 = 0$ vs $H_A: \text{not all of } \beta_4, \beta_5 = 0$ (both)

② Test statistic $F = \frac{MS_{\text{addition}}}{MSE(\text{model B})} \sim F_{2, 16}$

③ $\alpha = 0.05$ reject H_0 if $p < \alpha$

④ obs $F_{2, 16} = 4.6638$



⑤ As $p = 0.02536 < \alpha = 0.05$
 reject H_0 in favour of H_A & conclude that
 at least one of the two additional terms is
 a significant addition to the base model

The partial (nested) F test is the most general of these tests for nested models, but the differences in names here is just jargon!

A sequential F test is just the special case of a nested test where we are adding a single additional (they are both still partial F tests)

The overall F test for a multiple regression model is also just a special case of the nested F test \rightarrow here the additional terms are all of the x variables on top of a null model;

$$\text{Model (A)} : Y = \beta_0 + \varepsilon \quad \leftarrow \text{null or mean model}$$

$$\text{Model (B)} : Y = \beta_0 + \underbrace{\beta_1 x_1 + \beta_2 x_2 + \dots + \beta_k x_k}_{\text{addition!}} + \varepsilon$$

In all of these (partial) F tests for nested models, we will tend to prefer the simpler "base" model over the more complicated "expanded" model whenever we fail to reject H_0 (with some caveats \rightarrow the null model is not always the best "base" model \rightarrow it will depend on the research question)

\rightarrow but these F tests and the ANOVA table are key in deciding what belongs in the model, we will use these as an approach for refining models

What has changed from SLR to MR?

- `plot()` – is model appropriate?
are the underlying assumptions OK?
- these are pretty much the same as earlier
- `anova()` – is model adequate?
does it (and all parts of it)
have significant explanatory power?
- this has definitely changed
- `summary()` – used to try the research question,
once we have the right model
- again, much the same as earlier
- `predict()` – we will also see that this hasn't
changed much