

### A study of the relationship of body fat to several physical measurements was conducted on a sample of 20 healthy women aged 25-34 years. The predictor measurements were the triceps, skinfold, the thigh circumference and the midarm circumference. If we perform a multiple linear regression of body fat on the three predictors, the associated internally and externally Studentized residuals are:

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## None of these (internally or externally) studentised residuals seem excessively large. Suppose we wish to test whether the data point with the largest standardised residual is a mean-shift residual. We could compare the $t_i$ or $r_i$ value for the 14th observation with > bodyfat.lm\$df [1] 16 > qt(0.975, bodyfat.lm\$df) [1] 2.119905

## Bonferroni correction Five look at all $mt_1$ values, we are implicitly performing n hypothesis tests. We need to adjust $\alpha$ thresholds with Bonferoni An adjustment made to P values when several dependent or independent statistical tests are being performed simultaneously on a single data set. Divide $\alpha$ by the number of comparisons being made.

# Suppose a researcher is testing 20 hypotheses simultaneously, with a critical P value of 0.05. In this case, the following would be true: $P(\text{at least one significant result}) = 1 - P(nosignificant results) \\ P(\text{at least one significant result}) = 1 - (1 - 0.06)^{20} \\ P(\text{at least one significant result}) = 0.64 \\ So performing 20 tests on a data set yields a 64 percent chance of identifying at least one significant result, even if all of the tests are actually not significant. If you want an overall significance level <math>\alpha$ and you perform in individual tests, simply divide $\alpha$ by n to obtain the significance level required for the individual tests.

### Suppose alpha = 0.05, with 20 tests

> qt(1-0.05/(2\*20)), bodyfat.lm\$df) [1] 3.580522

Which is far larger than any of the studentised residual values