Lesson 10: Chapter 6 Sections 3-4

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BSC Mathematics

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Tests of significance have limitations, and it's important to understand these limits. In §6.3, we'll discuss

11 what levels of significance mean,

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- 5 when statistical inference is inappropriate to use, and
- 6 searching for significance.

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There's no cutoff for what is and is not significant. Our choice of α really is arbitrary. Obviously a P-value of 0.051 suggests that the data might be statistically significant, even though the level of significance might have been set at $\alpha=0.05$. But some data is obviously statistically significant. If the P-value for a test were 0.000001, then this result would arise from the underlying assumptions in the test only one in a million times! So the assumptions are almost certainly wrong. Sometimes a very small P-value is denoted by P<0.001.

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Often in very large samples, even small deviations from the null hypothesis will be significant. But this statistical significance doesn't mean that the significance is practical! In particular, strong significance doesn't imply dramatic difference from the null hypothesis. It just means that the results imply that our assumption was almost certainly wrong — but perhaps only a little wrong.

What Lack of Significance Means:

A study of a vaccine's effectiveness determined that 99% CI for the proportion change in number of infections was (-0.25,0.30). A newspaper reported that the vaccine didn't cause any change in the infection rate — because the null hypothesis in this case would have been that there is no effect from the vaccine and that case (where the proportion change is 0) is included in the 99% CI. What's wrong with this?

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Absence of evidence is not evidence of absence! In particular, there could be at most a 25% decrease or a 30% increase due to the vaccine (with 99% confidence). What really should be said is that the effectiveness of the vaccine has not been proven. Further study is necessary.

Statistical Inference Isn't Always Appropriate:

A study is interested in determining the veracity of the claim that 65% of voters in a statewide election are republicans, so they conduct an exit poll at a randomly selected polling station. They survey the 2581 voters who voted at the polling station and created a 99.99% CI for the proportion of registered republicans: (0.7500,0.7501). Does this support the claim?

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Searching for Significance:

Amazon mined its data and discovered that its customers who use Paypal to make purchases seemed to be more likely to make multiple purchases in the same month. Amazon sought verification of this hypothesis from a third party. Why shouldn't the third party use Amazon's data to test this hypothesis?

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Amazon mined its data and discovered that its customers who use Paypal to make purchases seemed to be more likely to make multiple purchases in the same month. Amazon sought verification of this hypothesis from a third party. Why shouldn't the third party use Amazon's data to test this hypothesis? You should never use data which suggests a hypothesis to verify the same hypothesis! Always conduct a new experiment/collect new data.

What does α really mean?

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Definition (type one error)

A **Type I Error** is the error of rejecting the null hypothesis when it is actually true — the probability of making a Type I Error is exactly α .

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You can improve power by increasing sample size, but it's not an easy task to compute the power of an α -level test.

If power measures how often you'll correctly reject H_0 when H_a is true, then obviously 1-(power) measures how often you'll support H_0 when H_a is true! We call this a Type II Error.

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Definition (type two error)

A **Type II Error** is the error of failing to reject the null hypothesis when the alternative hypothesis is actually true. If β is the power of a test, then the probability of a Type II Error is $1-\beta$.

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