

Malaria Molecular Surveillance Study Design Workshop

Module 3: Hypothesis testing and power



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In other cases, we have a specific question that we want to answer. This questions may be phrased as a **null hypothesis test**.



Sometimes we are simply trying to estimate something, e.g. prevalence. We have seen how to perform sample size calculation based on precision arguments.

In other cases, we have a specific question that we want to answer. This questions may be phrased as a **null hypothesis tests**.

A null hypothesis (H_0) is a statement of **no effect or difference** between groups. This is often a statement that nothing interesting is happening*

Rather than trying to prove there is an effect, in null hypothesis testing we try to **disprove** that there is **no effect**.

*Sometimes it can be very interesting if the null hypothesis is true



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• Q: Does vaccine efficacy vary based on genetic markers?



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 *H*₀: There is no association between genetic variant and gender or occupation.
- Q: Does vaccine efficacy vary based on genetic markers?



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- Q: Are certain genetic variants associated with gender, or occupation?
 *H*₀: There is no association between genetic variant and gender or occupation.
- Q: Does vaccine efficacy vary based on genetic markers?
 H₀: Vaccine efficacy is the same irrespective of genetic markers.



Each test has a **test statistic**

One-sample z-test for proportions: tests prevalence against a known value



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One-sample z-test for proportions: tests prevalence against a known value H_0 : The population prevalence equals p_0

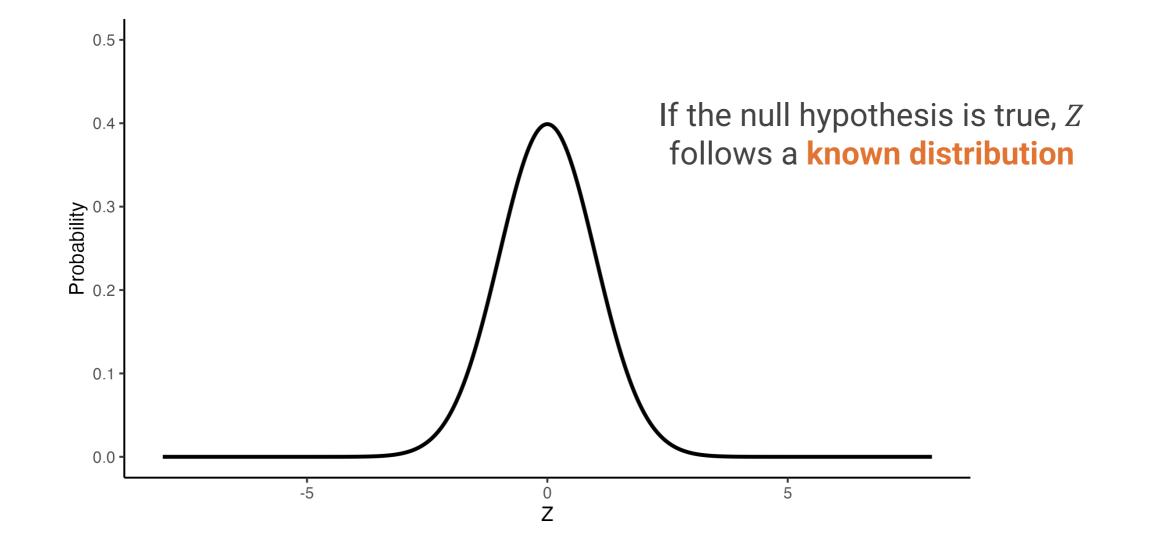


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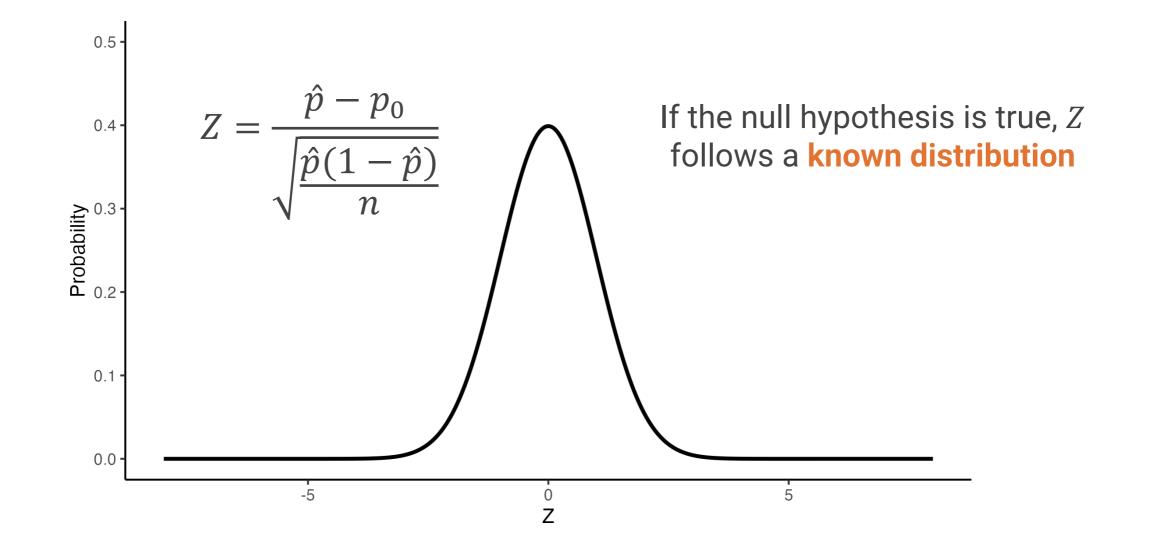
One-sample z-test for proportions: tests prevalence against a known value H_0 : The population prevalence equals p_0

$$Z = \frac{\hat{p} - p_0}{\sqrt{\frac{\hat{p}(1-\hat{p})}{n}}}$$

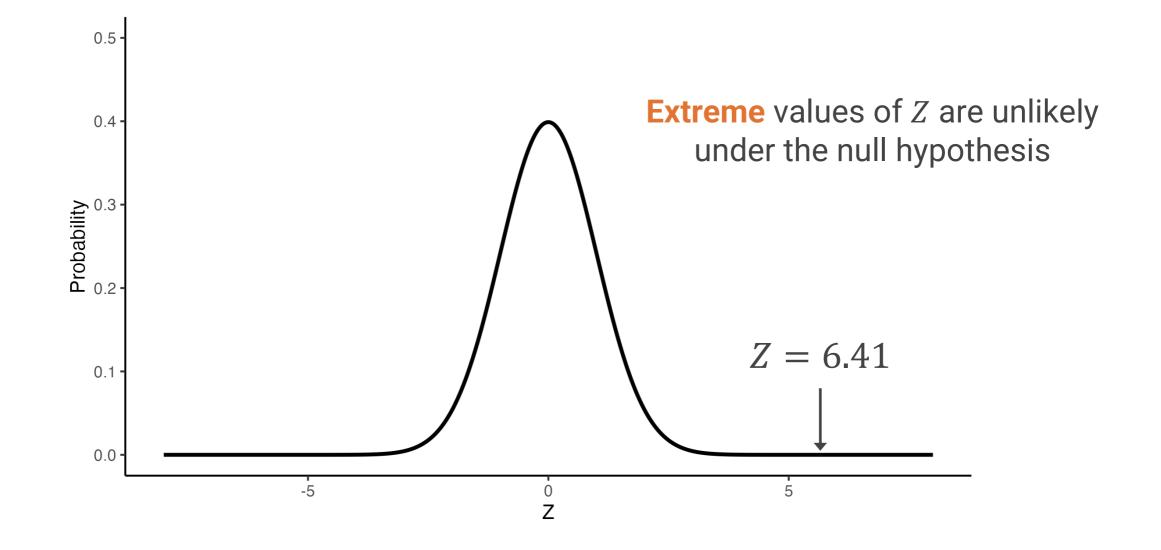




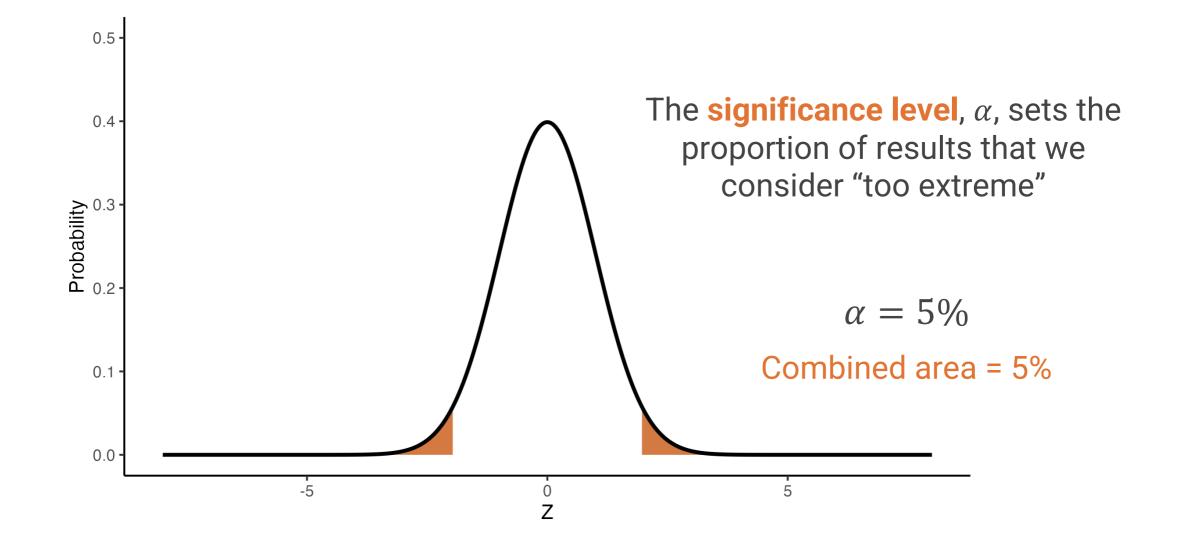




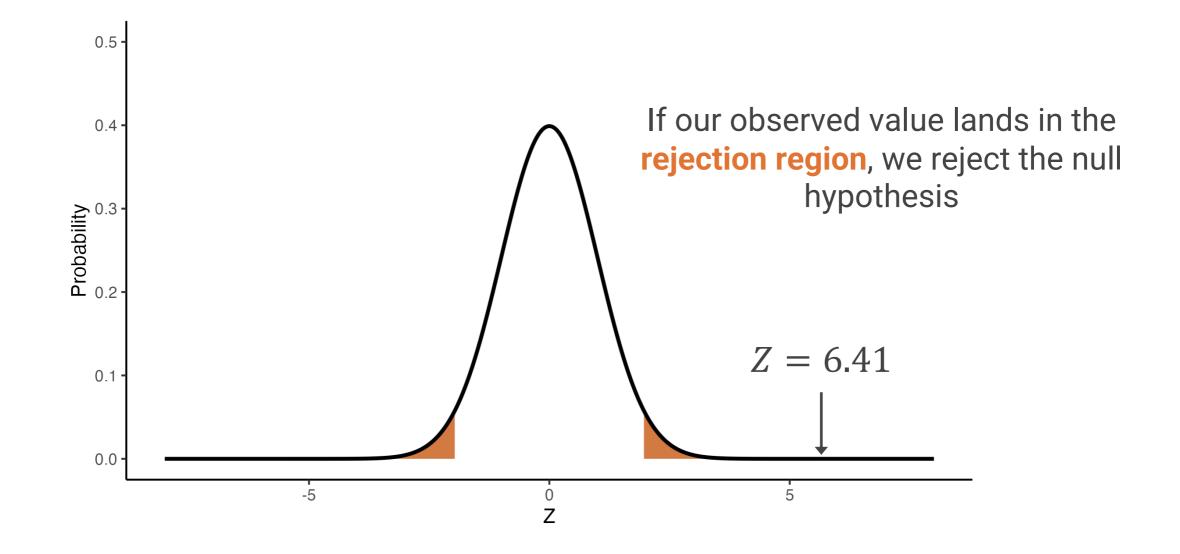




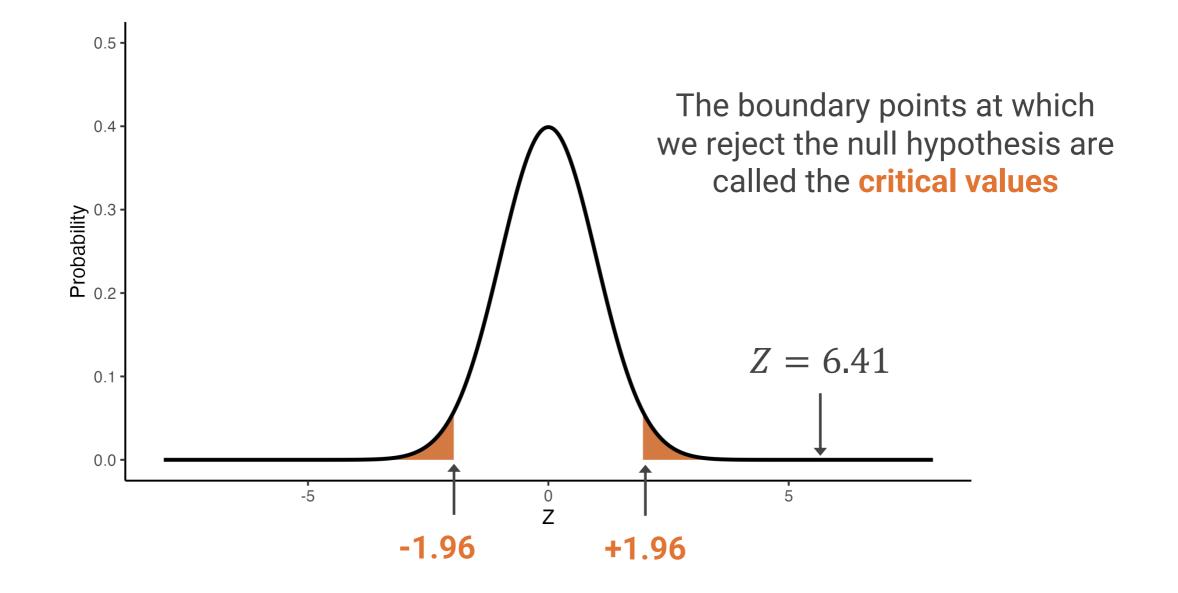




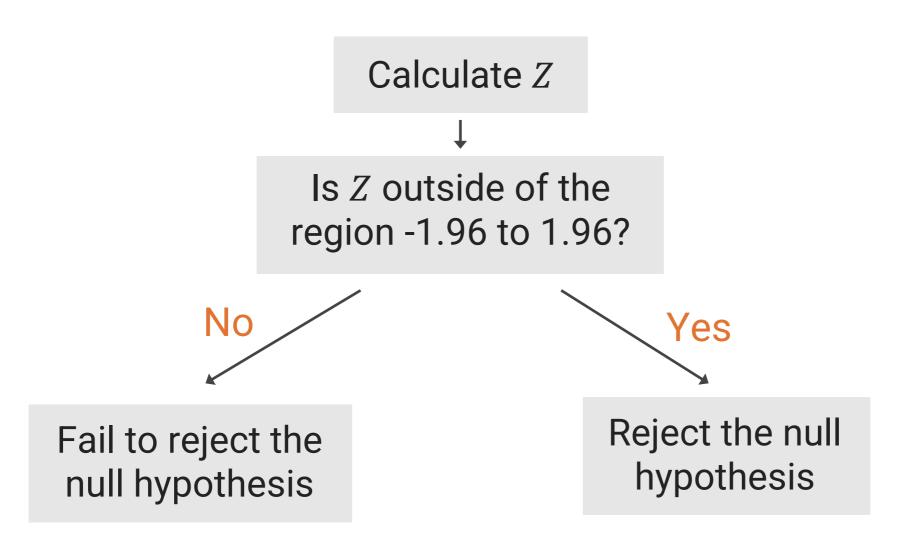














		Conclusion about <i>H</i> ₀	
		Fail to reject	Reject
Truth about <i>H</i> ₀	True	True negative	False positive
	False		



		Conclusion about <i>H</i> ₀	
		Fail to reject	Reject
Truth	True	True negative $1 - \alpha$	False positive α
about <i>H</i> ₀	False		



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 α sets the **false positive rate** of a test. Using α we can control how often we incorrectly conclude that there is a real effect when there is none.



		Conclusion about <i>H</i> ₀	
		Fail to reject	Reject
Truth about <i>H</i> ₀	True	True negative $1 - \alpha$	False positive α
	False	What about this!?	

 α sets the **false positive rate** of a test. Using α we can control how often we incorrectly conclude that there is a real effect when there is none.



In power analysis, we also specify an **alternative hypothesis**

- H_0 : The population prevalence equals p_0
- H_1 : The population prevalence equals p, which is different from p_0



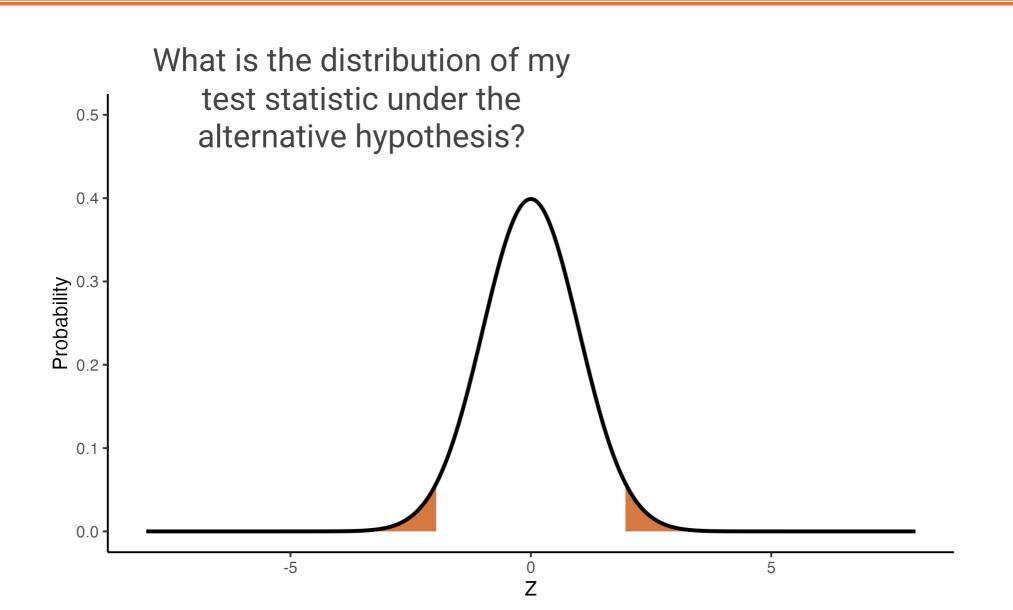
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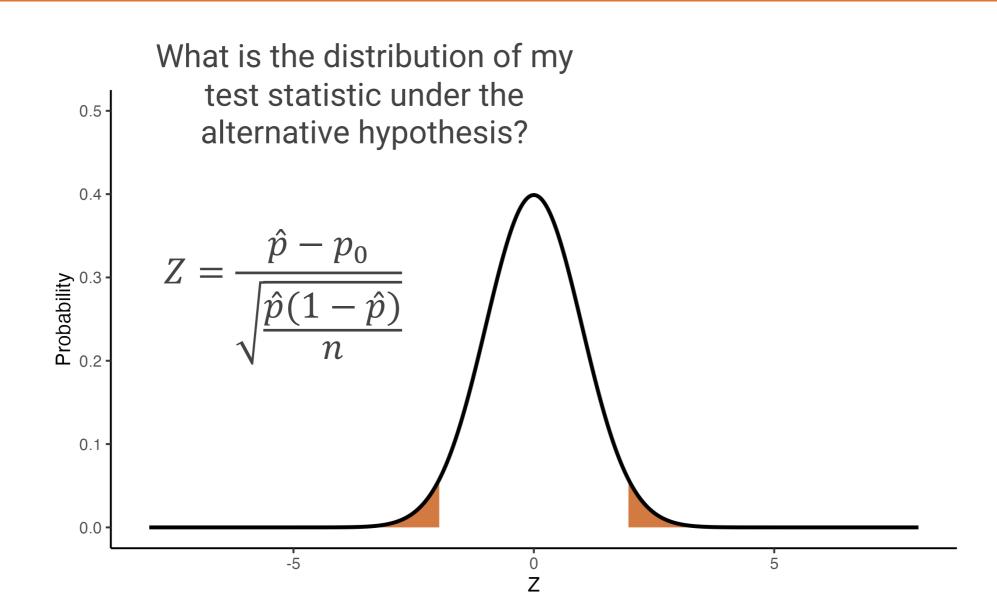
For example...

I want to test if the prevalence of *pfcrt* K76T mutations is significantly different from 10%. When powering this test, I assume the true prevalence of K76T mutations is 15%.

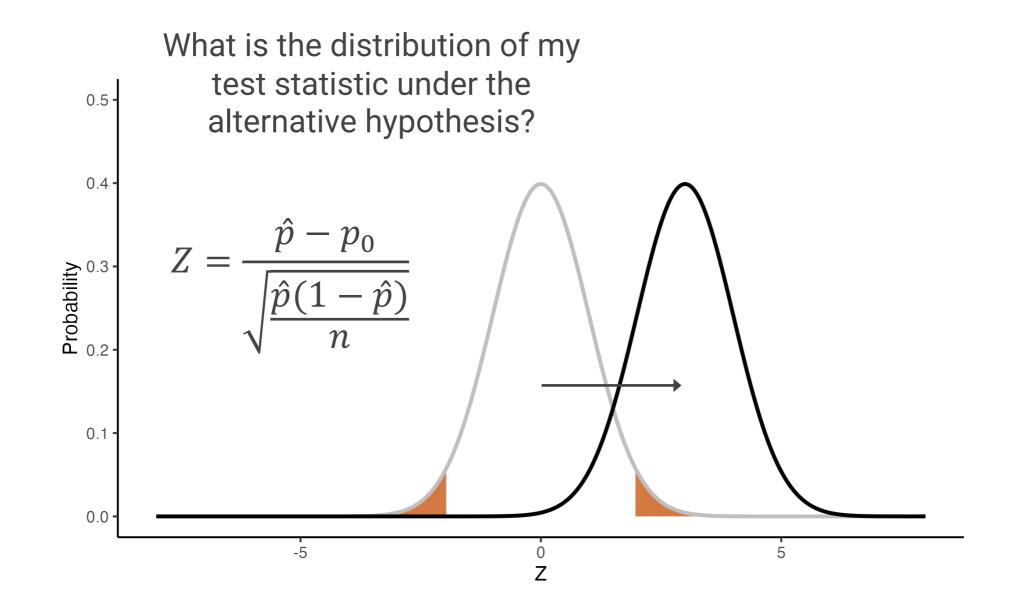




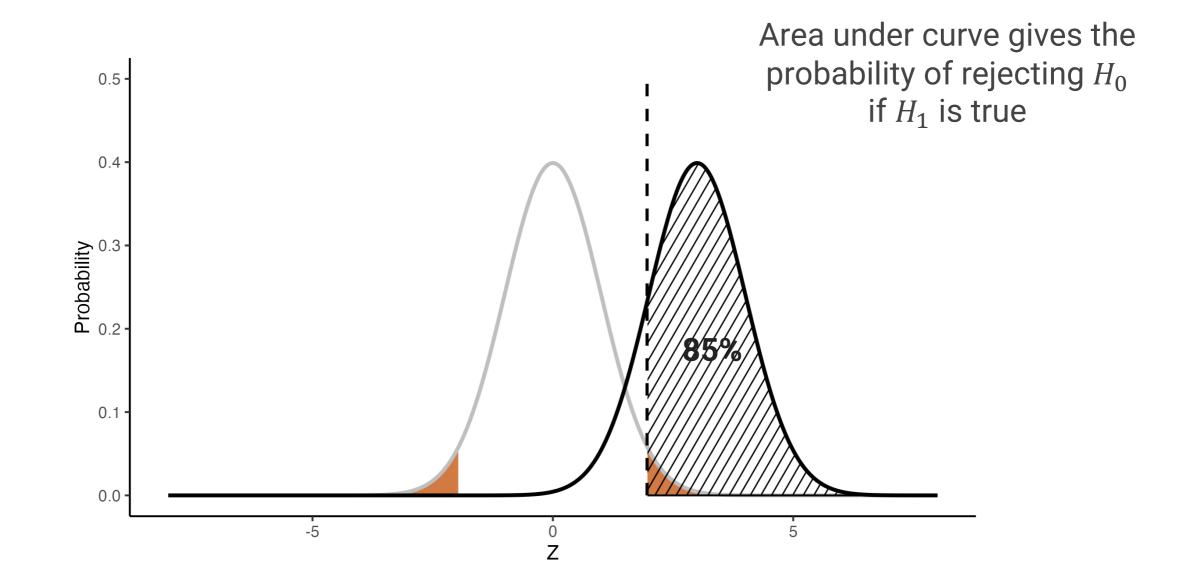




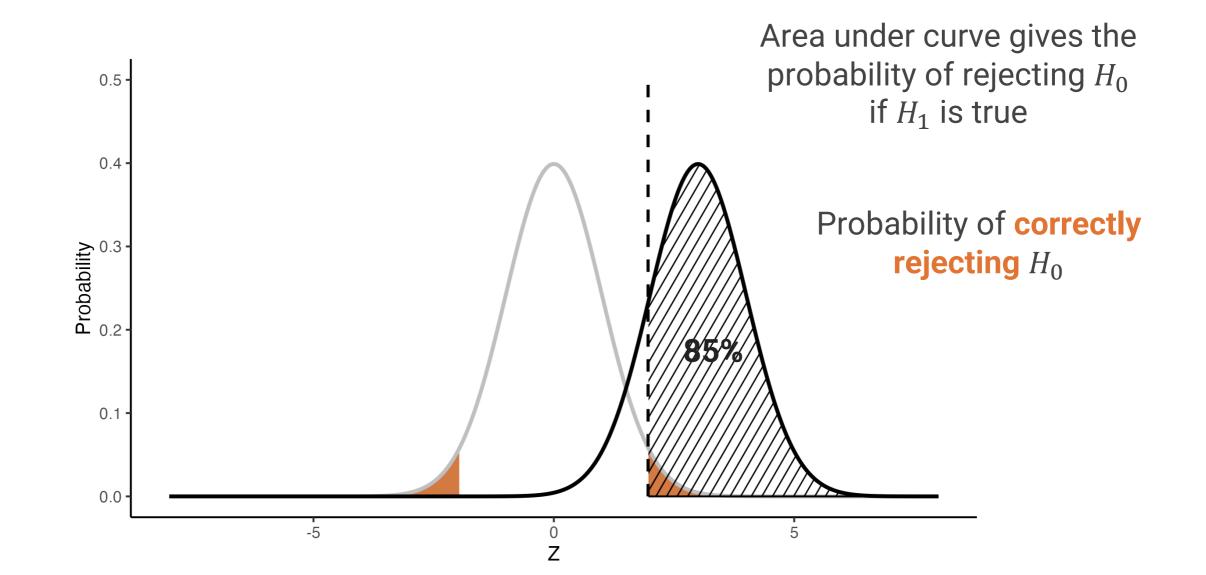




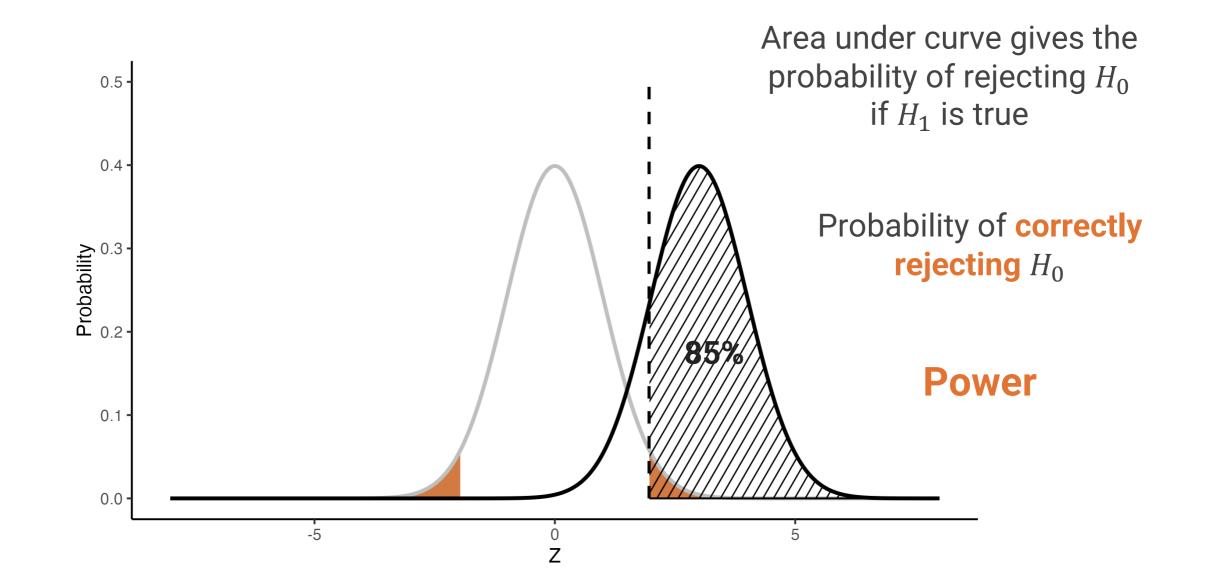














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	False	False negative	True positive



		Conclusion about <i>H</i> ₀	
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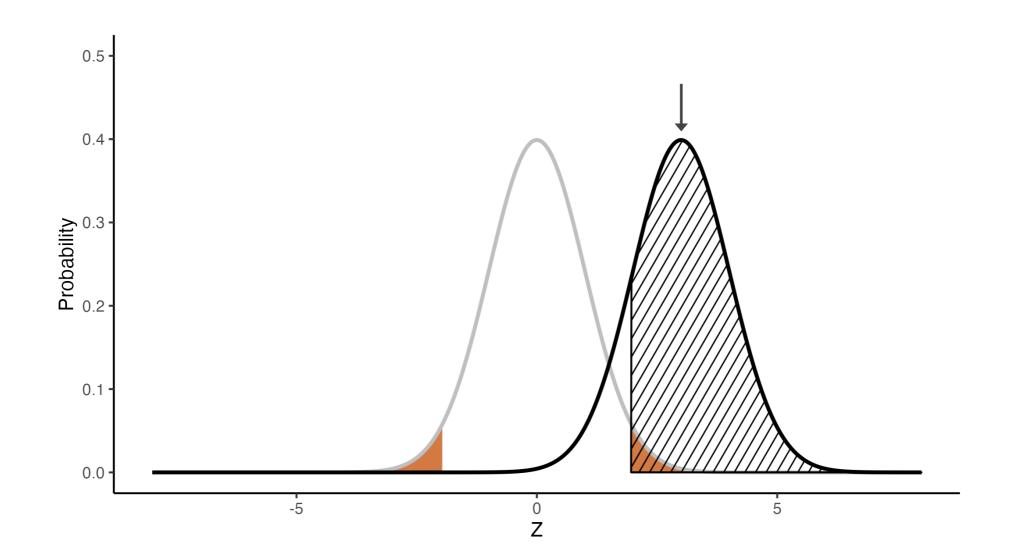


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		Fail to reject	Reject
Truth about <i>H</i> ₀	True	True negative $1 - \alpha$	False positive α
	False	False negative 1 – Power	True positive Power

Power is the probability of **correctly rejecting** the null hypothesis. It is the chance that we find something interesting, given that it is there.

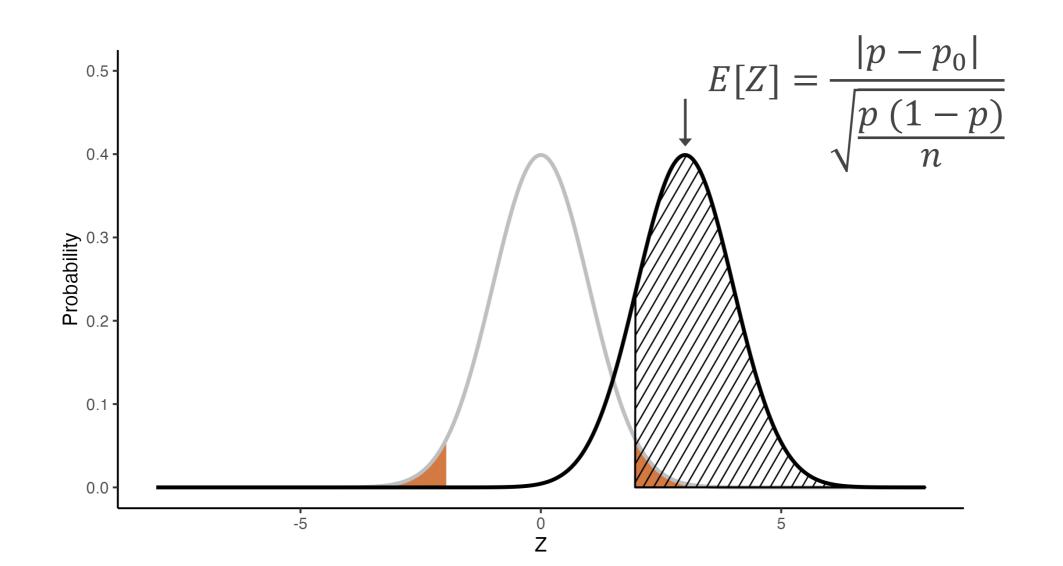
How do we calculate power?





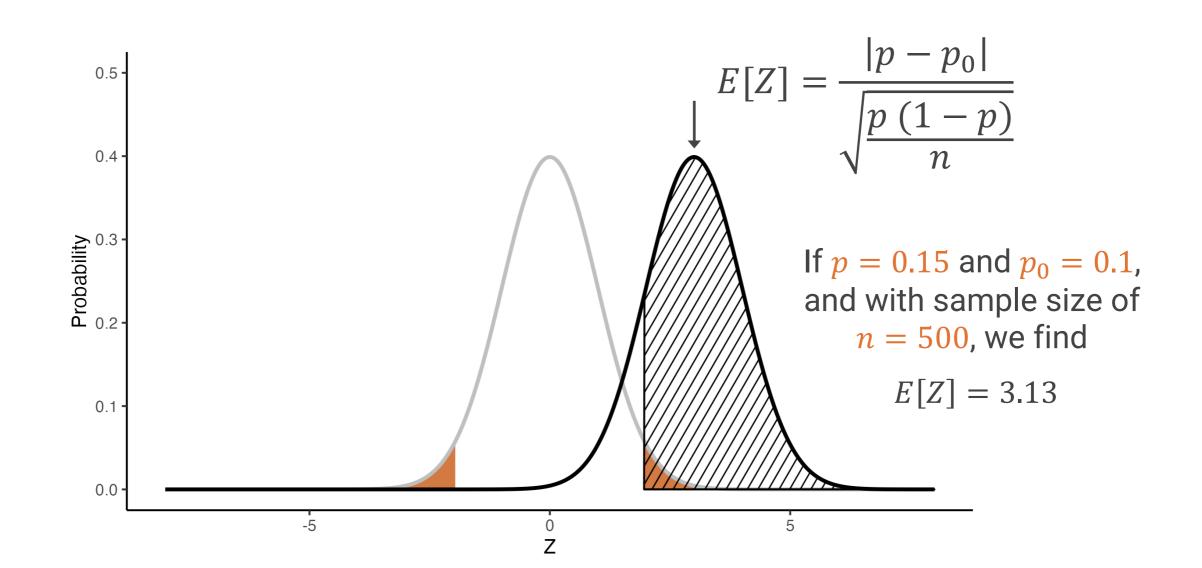
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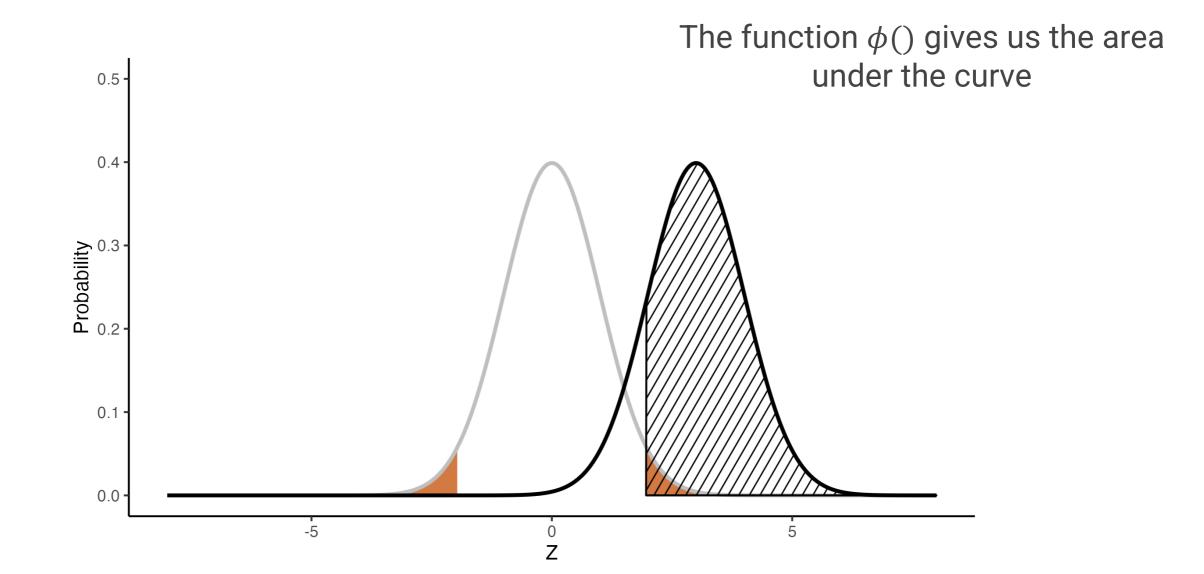


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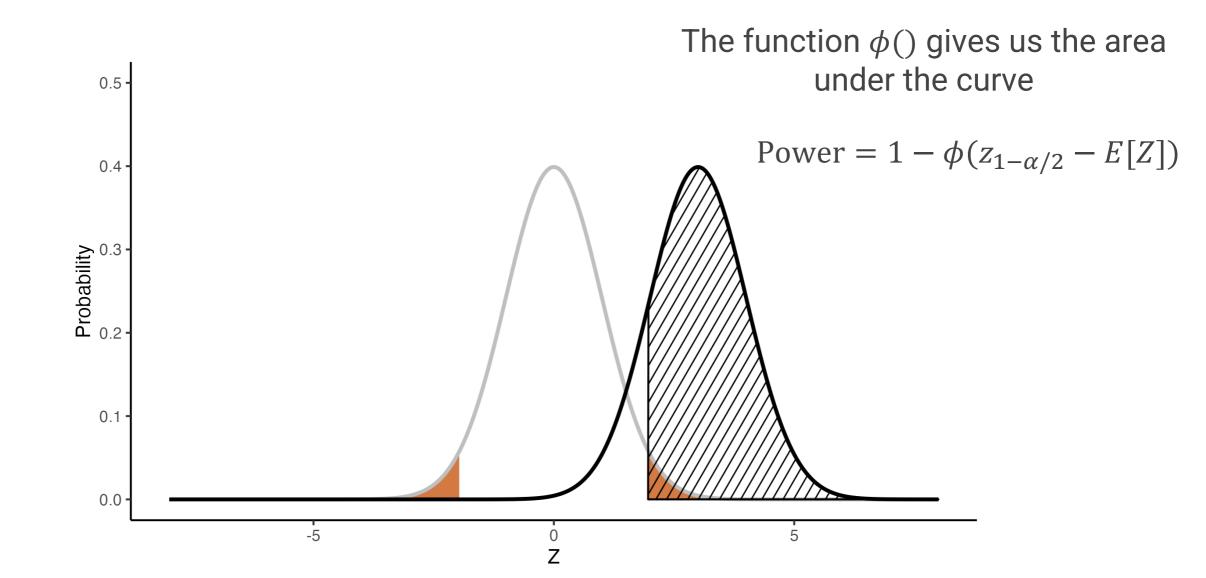




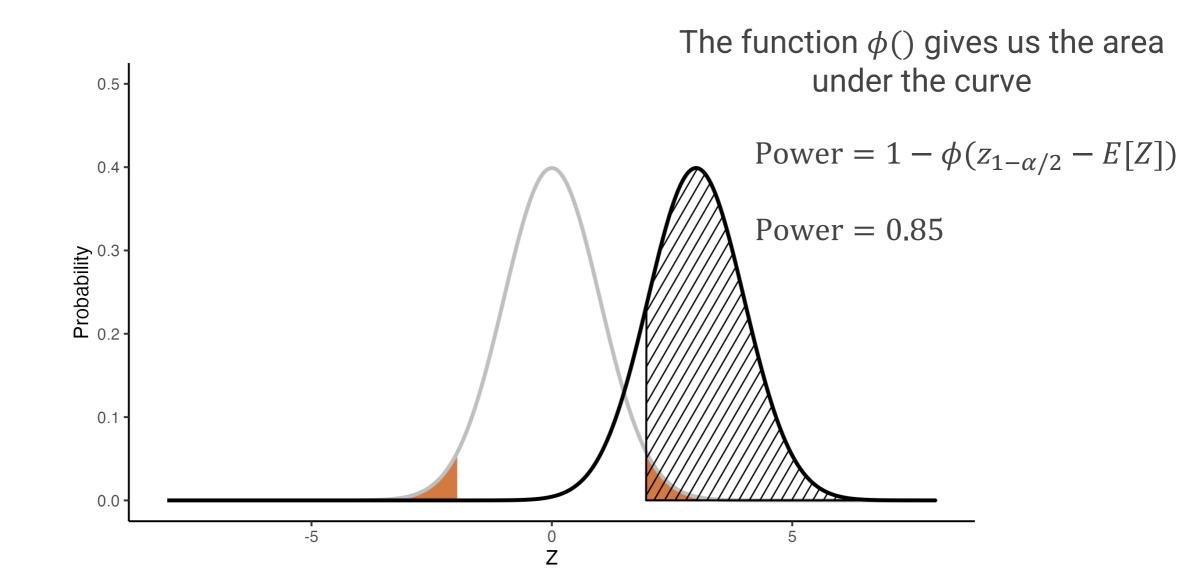














Power =
$$1 - \phi(z_{1-\alpha/2} - E[Z])$$

Power as a function of sample size



Power =
$$1 - \phi \left(z_{1-\alpha/2} - \frac{|p - p_0|}{\sqrt{\frac{p(1-p)}{n}}} \right)$$

Power as a function of sample size

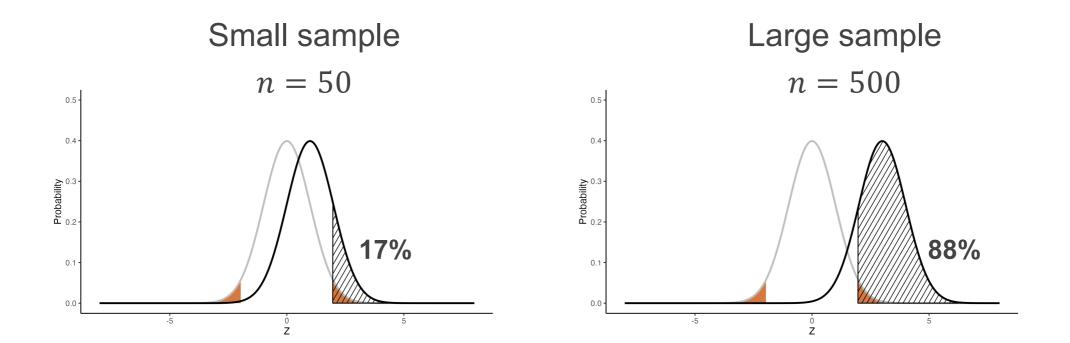


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 Power varies as a function of sample size

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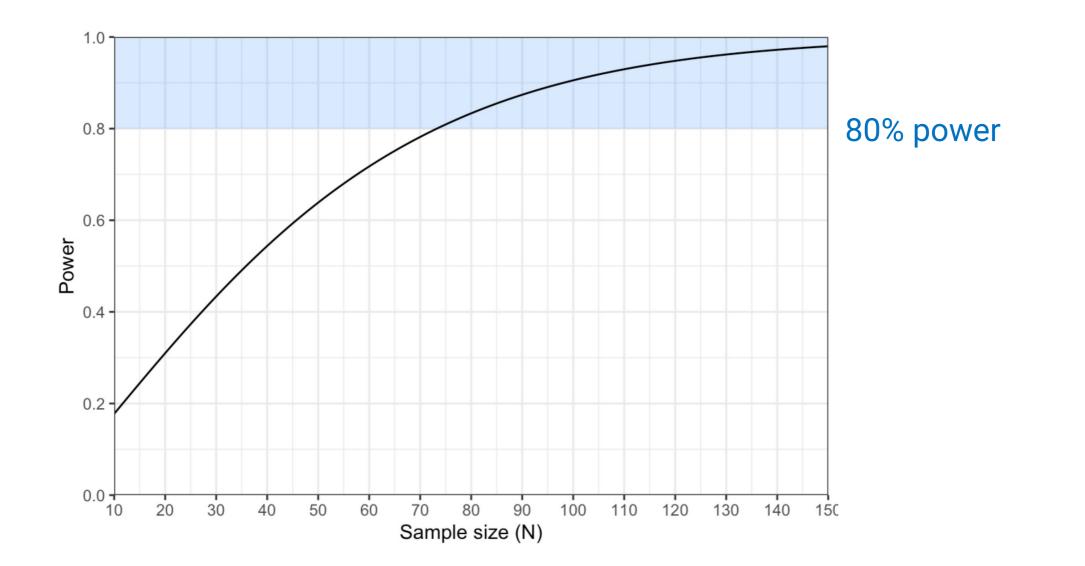


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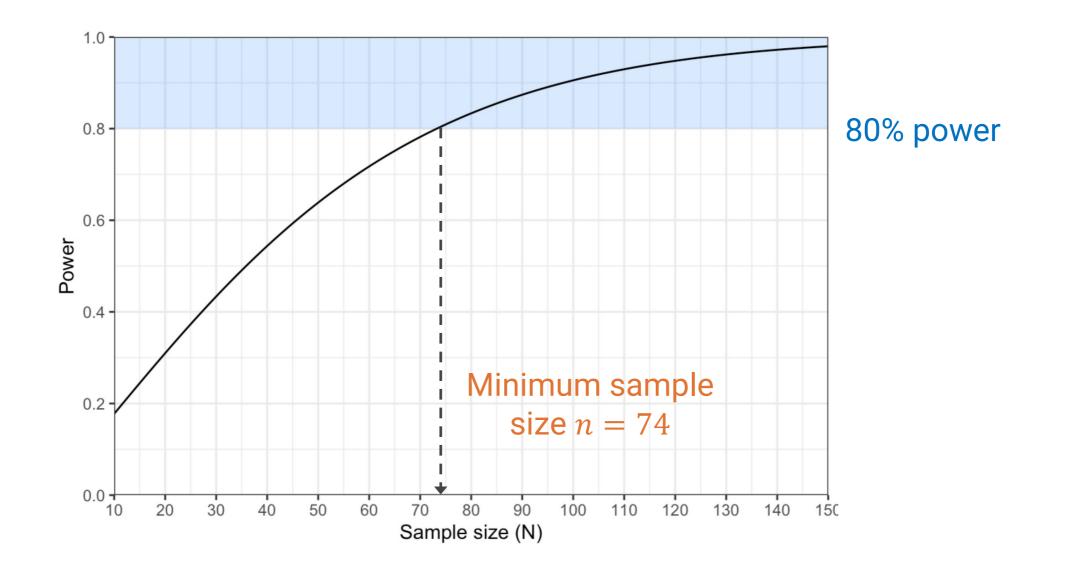
Power curves





Power curves





Sample size formulae



Power =
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Can we reverse-engineer this to find the value of *n* that achieves a target power?

Sample size formulae



Power =
$$1 - \phi \left(z_{1-\alpha/2} - \frac{|p - p_0|}{\sqrt{\frac{p(1-p)}{n}}} \right)$$

Can we reverse-engineer this to find the value of *n* that achieves a target power?

$$n = \left(z_{1-\beta} + z_{1-\frac{\alpha}{2}}\right)^2 \frac{p(1-p)}{(p-p_0)^2}$$

Where $\beta = 1 - Power$. For 80% power, we find $z_{1-\beta} = 0.84$



- We can ask questions using **null hypothesis tests**
- A null hypothesis is a statement of no effect/difference between groups
- The significance level α controls the **false-positive rate**
- Power is the true positive rate. It is the chance of correctly rejecting the null hypothesis.
- Power increases with sample size. We can use power curves or sample size formulae to choose a value of n



Format: Interactive R code, accessed through the web

- Short quiz on hypothesis testing
- Test for change in prevalence
 - Calculate power
 - Calculate minimum sample size
- Test for detection of rare *pfk13* variant
 - Calculate power
 - Calculate minimum sample size



https://tinyurl.com/bd4um5mj