

# **Malaria Molecular Surveillance Study Design Workshop**

# **Module 3:** Hypothesis testing and power



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 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



• Q: Are certain genetic variants associated with gender, or occupation?

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



 $H_0$ : Prevalence has remained the same over the last 5 years.

• Q: Are certain genetic variants associated with gender, or occupation?

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



 $H_0$ : Prevalence has remained the same over the last 5 years.

- 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?



 $H_0$ : Prevalence has remained the same over the last 5 years.

- 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?  $H_0$ : 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



Each test has a **test statistic**

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



Each test has a **test statistic**

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

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









































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





 $\alpha$  sets the **false positive rate** of a test. Using  $\alpha$  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**

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



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

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

### **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%.









































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



























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

## **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**



Power = 
$$
1 - \phi \left( z_{1-\alpha/2} - \frac{|p-p_0|}{\sqrt{\frac{p(1-p)}{n+1}}} \right)
$$
 Power varies as a function of sample size

#### **Power as a function of sample size**



Power = 
$$
1 - \phi \left( z_{1-\alpha/2} - \frac{|p-p_0|}{\sqrt{\frac{p(1-p)}{n+1}}}\right)
$$
 Power varies as a function of sample size



#### **Power curves**





#### **Power curves**





#### **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?

#### **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?

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

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  $\alpha$  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