

## Lecture 16: Wald tests

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## Wald test for one parameter

## Wald test for one parameter: examples

- ▶ Population mean:  $H_0 : \mu = \mu_0$
- ▶ Population proportion:  $H_0 : p = p_0$
- ▶ Regression coefficient:  $H_0 : \beta_j = 0$

## Testing multiple parameters

Logistic regression model for the dengue data:

$$Y_i \sim \text{Bernoulli}(p_i)$$

$$\log \left( \frac{p_i}{1 - p_i} \right) = \beta_0 + \beta_1 WBC_i + \beta_2 PLT_i$$

Researchers want to know if there is any relationship between white blood cell count or platelet count, and the probability a patient has dengue.

**Question:** What hypotheses should they test?

## Testing multiple parameters

##	Estimate	Std. Error	z value	Pr(> z )
## (Intercept)	2.642	0.121	21.772	0
## WBC	-0.289	0.013	-21.533	0
## PLT	-0.007	0.001	-11.061	0

$$H_0 : \beta_1 = \beta_2 = 0$$

Can the researchers test their hypotheses using this output?

## Wald tests for multiple parameters

For the dengue example:

$$\hat{\beta} = \begin{pmatrix} \hat{\beta}_0 \\ \hat{\beta}_1 \\ \hat{\beta}_2 \end{pmatrix}$$

```
m1 <- glm(Dengue ~ WBC + PLT, data = dengue, family = binom)
coef(m1)
```

```
## (Intercept)          WBC          PLT
## 2.641506279 -0.289290446 -0.006561464
```

```
vcov(m1)
```

```
##          (Intercept)          WBC          PLT
## (Intercept) 1.471934e-02 -4.937020e-04 -5.125888e-05
## WBC         -4.937020e-04  1.804972e-04 -3.221337e-06
## PLT         -5.125888e-05 -3.221337e-06  3.518938e-07
```

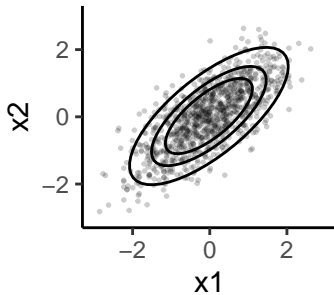
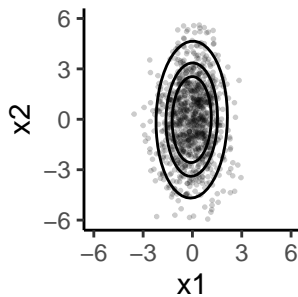
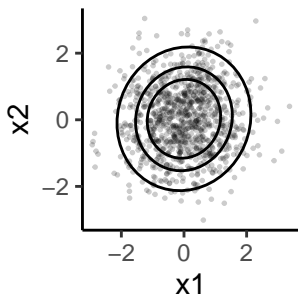
# Multivariate normal distribution

**Definition:** Let  $X = (X_1, \dots, X_k)^T$ . We say that  $X \sim N(\mu, \Sigma)$  if for any  $\mathbf{a} \in \mathbb{R}^k$ ,  $\mathbf{a}^T X$  follows a (univariate) normal distribution.

►  $\mu =$

►  $\Sigma =$

## Multivariate normal distribution





## Wald tests for multiple parameters

For the dengue example:  $\hat{\beta} = \begin{pmatrix} \hat{\beta}_0 \\ \hat{\beta}_1 \\ \hat{\beta}_2 \end{pmatrix}$

We want to test:  $\begin{pmatrix} \beta_1 \\ \beta_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$

## Wald tests for multiple parameters

$$H_0 : \mathbf{C}\beta = \gamma_0$$

For the dengue example:

```
C <- matrix(c(0, 1, 0,  
              0, 0, 1), nrow=2, byrow=T)
```

```
C
```

```
##      [,1] [,2] [,3]  
## [1,]    0    1    0  
## [2,]    0    0    1
```

```
C %*% coef(m1)
```

```
##      [,1]  
## [1,] -0.289290446  
## [2,] -0.006561464
```

## Wald tests for multiple parameters

For the dengue example:

```
C
```

```
##           [,1] [,2] [,3]
## [1,]         0    1    0
## [2,]         0    0    1
```

```
vcov(m1)
```

```
##              (Intercept)              WBC              PLT
## (Intercept)  1.471934e-02 -4.937020e-04 -5.125888e-05
## WBC          -4.937020e-04  1.804972e-04 -3.221337e-06
## PLT          -5.125888e-05 -3.221337e-06  3.518938e-07
```

```
C %*% vcov(m1) %*% t(C)
```

```
##              [,1]          [,2]
## [1,]  1.804972e-04 -3.221337e-06
## [2,] -3.221337e-06  3.518938e-07
```

## Wald tests for multiple parameters

►  $H_0 : \mathbf{C}\beta = \gamma_0$

► Look at  $\mathbf{C}\hat{\beta}$

**Fact:** Suppose  $X \sim N(\mu, \Sigma)$  (multivariate normal), and  $\mathbf{A}$  is a matrix (not random). Then:

$$\mathbf{A}X \sim$$

## Test statistic and p-value

- ▶  $H_0 : \mathbf{C}\beta = \gamma_0$
- ▶  $\mathbf{C}\hat{\beta} \approx N(\mathbf{C}\beta, \mathbf{C}\mathcal{I}^{-1}(\beta)\mathbf{C}^T)$
- ▶ Want to turn  $\mathbf{C}\hat{\beta}$  into a scalar test statistic

## Example

$$H_0 : \mathbf{C}\beta = \gamma_0$$

$$(\mathbf{C}\hat{\beta} - \gamma_0)^T (\mathbf{C}\mathcal{I}^{-1}(\beta)\mathbf{C}^T)^{-1} (\mathbf{C}\hat{\beta} - \gamma_0) \approx \chi_q^2 \quad \text{under } H_0$$

```
C %*% coef(m1)

## [1] -0.289290446 -0.006561464

test_stat <- t(C %*% coef(m1)) %*% solve(C %*% vcov(m1) %*%
  (C %*% coef(m1)))
test_stat

##           [,1]
## [1,] 930.5777

pchisq(test_stat, 2, lower.tail=F)

##           [,1]
## [1,] 8.464858e-203
```