

# Logistic regression

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## Last time: Dengue data

**Data:** Data on 5720 Vietnamese children, admitted to the hospital with possible dengue fever. Variables include:

- ▶ *Sex*: patient's sex (female or male)
- ▶ *Age*: patient's age (in years)
- ▶ *WBC*: white blood cell count
- ▶ *PLT*: platelet count
- ▶ other diagnostic variables...
- ▶ *Dengue*: whether the patient has dengue (0 = no, 1 = yes)

**Research goal:** Predict dengue status using diagnostic measurements

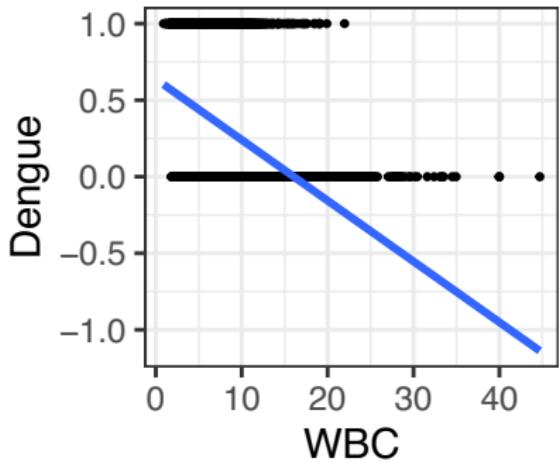
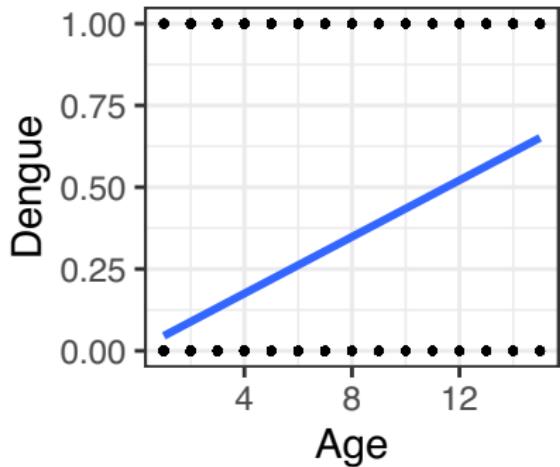
## Last time: initial attempt

What if we try a linear regression model?

$Y_i$  = dengue status of  $i$ th patient

$$Y_i = \beta_0 + \beta_1 WBC_i + \varepsilon_i \quad \varepsilon_i \stackrel{iid}{\sim} N(0, \sigma^2)$$

## Don't fit linear regression with a binary response



## Last time: rewriting the linear regression model

$$Y_i|WBC_i \sim N(\mu_i, \sigma^2) \quad (\text{random component})$$

$$\mu_i = \beta_0 + \beta_1 WBC_i \quad (\text{systematic component})$$

↑  
specifies new parameters  
of the distribution  
depends on wbc

## Second attempt

$$Y_i | WBC_i \sim \text{Bernoulli}(p_i) \quad p_i = \mathbb{P}(Y_i = 1 | WBC_i)$$

$$p_i = \beta_0 + \beta_1 WBC_i$$

Are there still any potential issues with this approach?

Problem :  $p_i \in [0, 1]$

but  $\beta_0 + \beta_1 WBC_i \in (-\infty, \infty)$   
(potentially)

## Fixing the issue: logistic regression

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

(random component)

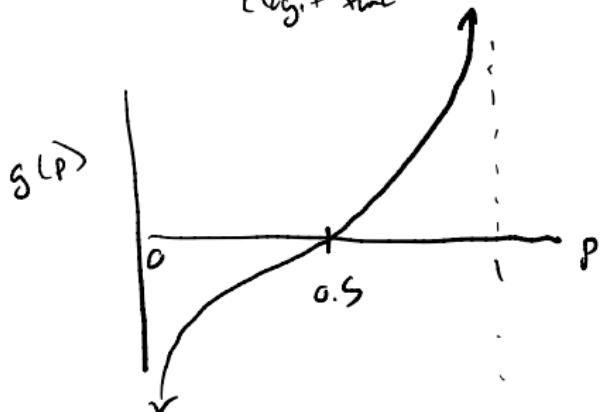
link function  $\rightsquigarrow g(p_i) = \beta_0 + \beta_1 WBC_i$

(systematic component)

where  $g : (0, 1) \rightarrow \mathbb{R}$  is unbounded.

**Usual choice:**  $g(p_i) = \log\left(\frac{p_i}{1-p_i}\right)$

(logit function)



$$p_i \in (0, 1)$$

$$\frac{p_i}{1-p_i} = \frac{\text{odds}}{\in (0, \infty)}$$

$$\log\left(\frac{p_i}{1-p_i}\right) = \text{log odds} \in (-\infty, \infty)$$

## Logistic regression model

$$Y_i | WBC_i \sim \text{Bernoulli}(p_i) \quad \leftarrow \begin{matrix} \text{capturing randomness} \\ \text{about } Y_i \end{matrix}$$

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

Why is there no noise term  $\varepsilon_i$  in the logistic regression model?

Discuss for 1–2 minutes with your neighbor, then we will discuss as a class.

Linear model:

$$Y_i = \beta_0 + \beta_1 X_i + \varepsilon_i; \quad \varepsilon_i \sim N(0, \sigma^2)$$

$\hookrightarrow$

$$Y_i | X_i \sim N(\mu_i, \sigma^2)$$

$$\mu_i = \beta_0 + \beta_1 X_i$$

## Fitting the logistic regression model

$$Y_i | WBC_i \sim Bernoulli(p_i)$$

"generalized linear model"

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

```
m1 <- glm(Dengue ~ WBC, data = dengue,  
            family = binomial)  
summary(m1)
```

family  
of  
specifies distribution  
the response

## Fitting the logistic regression model

$$Y_i | WBC_i \sim Bernoulli(p_i)$$

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

```
m1 <- glm(Dengue ~ WBC, data = dengue,  
            family = binomial)
```

```
summary(m1)
```

```
##
```

```
## Call:
```

```
## glm(formula = Dengue ~ WBC, family = binomial, data = de
```

```
##
```

```
## Coefficients:
```

	Estimate	Std. Error	z value	Pr(> z )
## (Intercept)	1.73743	0.08499	20.44	<2e-16 ***
## WBC	-0.36085	0.01243	-29.03	<2e-16 ***
## ---				

$$\log\left(\frac{\hat{p}_i}{1-\hat{p}_i}\right) = 1.737 - 0.361WBC_i$$



*z-values instead of t-values*

*t-values*

## Activity

Work on the activity (handout) in groups, then we will discuss as a class.

## Activity

$$\log \left( \frac{\hat{p}_i}{1 - \hat{p}_i} \right) = 1.737 - 0.361 \ WBC_i$$

What is the predicted *odds* of dengue for a patient with a white blood cell count of 15?

$$\begin{aligned}\hat{\text{log odds}} &= 1.737 - 0.361(15) \\ &= -3.678\end{aligned}$$

$$\begin{aligned}\hat{\text{odds}} &= e^{-3.678} \\ &= 0.0253\end{aligned}$$

## Activity

$$\log \left( \frac{\hat{p}_i}{1 - \hat{p}_i} \right) = 1.737 - 0.361 WBC_i$$

What is the predicted *probability* of dengue for a patient with a WBC of 15?

$$\text{odds} = \frac{P}{1-P} \quad \Leftrightarrow \quad P = \frac{\text{odds}}{1 + \text{odds}}$$
$$= \frac{e^{\text{log odds}}}{1 + e^{\text{log odds}}}$$

$$\hat{p}_i = \frac{e^{-3.678}}{1 + e^{-3.678}} \approx 0.025$$

## Interpretation: Activity 2

Work on the activity (handout) in groups, then we will discuss as a class.

## Interpretation

$$\log \left( \frac{\hat{p}_i}{1 - \hat{p}_i} \right) = 1.737 - 0.361 WBC_i$$

Are patients with a higher WBC more or less likely to have dengue?

less likely      (negative slope)  
 $\hat{\beta}_1 < 0$

## Interpretation

$$\log \left( \frac{\hat{p}_i}{1 - \hat{p}_i} \right) = 1.737 - 0.361 \ WBC_i$$

What is the change in *log odds* associated with a unit increase in WBC?

-0.361

## Interpretation

$$\log \left( \frac{\hat{p}_i}{1 - \hat{p}_i} \right) = 1.737 - 0.361 WBC_i$$

What is the change in *odds* associated with a unit increase in WBC?

$$\begin{aligned} \frac{\text{odds when } WBC=x+1}{\text{odds when } WBC=x} &= \frac{e^{1.737 - 0.361(x+1)}}{e^{1.737 - 0.361x}} \\ &= e^{-0.361} \\ &= 0.697 \end{aligned}$$

A one-unit increase in WBC is associated with a change in the odds of dengue by a factor of 0.697 (30.3% decrease)

## Coefficient interpretation

$$\log \left( \frac{\hat{p}_i}{1 - \hat{p}_i} \right) = \hat{\beta}_0 + \hat{\beta}_1 X_i$$

## Fitting a *logistic* regression model?

Linear regression: minimize  $\sum_{i=1}^n (Y_i - \beta_0 - \beta_1 X_{i1} - \cdots - \beta_k X_{ik})^2$

**Question:** Should we minimize a similar sum of squares for a *logistic* regression model?