

Introduction to Logistic Regression

Agenda

- + Introductions
- + Overview of course details
- + Begin logistic regression
- + HW1 released on course website

Class overview

- + STA 711 focuses on *statistical inference*: estimation, confidence intervals, and hypothesis testing
- + Throughout the semester, topics will be initially motivated by logistic regression
- + We will continue with inference and GLMS in STA 712 (Generalized Linear Models)

Grading philosophy

- + Focusing on grades can detract from the learning process
- + Homework should be an opportunity to *practice* the material. It is ok to make mistakes when practicing, as long as you make an honest effort
- + Errors are a good opportunity to learn and revise your work
- + Partial credit and weighted averages of scores make the meaning of a grade confusing. Does an 85 in the course mean you know 85% of everything, or everything about 85% of the material?

Grading in this course

- + I will give you feedback on every assignment
- + All assignments are graded as Mastered / Not yet mastered
- + If you haven't yet mastered something, you get to try again!

Course components

- + Regular homework assignments
 - + Practice material from class
 - + A subset of questions will be graded
 - + You may resubmit "Not yet mastered" questions once
- + 3 take-home exams
 - + Opportunity to demonstrate mastery of course material
 - + Optional make-up exams for "Not yet mastered" questions
- + Optional final exam
 - + Final opportunity to demonstrate mastery

Assigning grades

To get a **C** in the course:

- + Receive credit for at least 4 homework assignments
- + Master at least 80% of the questions on one exam

To get a **B** in the course:

- + Receive credit for at least 5 homework assignments
- + Master at least 80% of the questions on two exams

To get an **A** in the course:

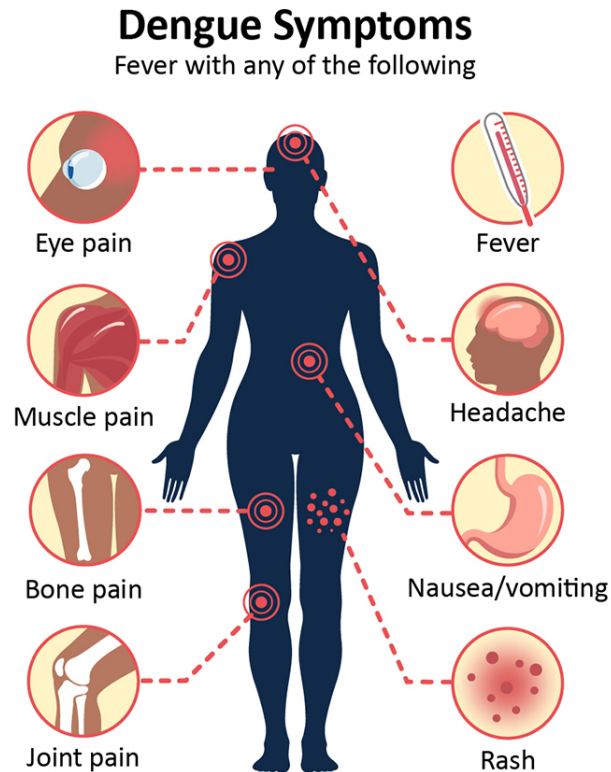
- + Receive credit for at least 5 homework assignments
- + Master at least 80% of the questions on all three exams

Late work and resubmissions

- + You get a bank of **5** extension days. You can use 1--2 days on any assignment, exam, or project.
- + No other late work will be accepted (except in extenuating circumstances!)

Motivating example: Dengue fever

Dengue fever: a mosquito-borne viral disease affecting 400 million people a year



Motivating example: 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)

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Research questions:

- + How well can we predict whether a patient has dengue?
- + Which diagnostic measurements are most useful?
- + Is there a significant relationship between WBC and dengue?

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- + How well can we predict whether a patient has dengue?
- + Which diagnostic measurements are most useful?
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How can I answer each of these questions? Discuss with a neighbor for 2 minutes, then we will discuss as a class.

- EDA (plots of dengue vs. WBC, dengue vs. Age, etc...)
 - ask experts / clients for context
- Fit regression model (e.g., logistic regression)
- CIs, hyp. tests, effect sizes for coefficients
- Model selection (stepwise selection, penalized, etc.)
- Prediction metrics (confusion matrix, accuracy, etc.)
 - cross validation

Fitting a model: initial attempt

What if we try a linear regression model?

Y_i = dengue status of i th patient

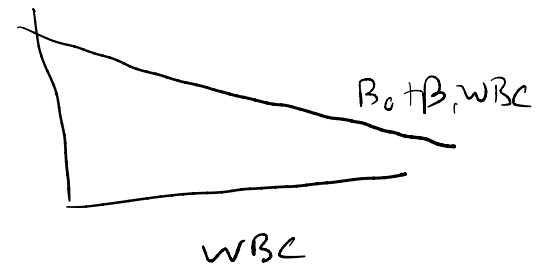
$$Y_i = \beta_0 + \beta_1 WBC_i + \varepsilon_i \quad \varepsilon_i \stackrel{iid}{\sim} N(0, \sigma_\varepsilon^2)$$

What are some potential issues with this linear regression model?

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

and is continuous

but Y_i is binary!



Second attempt

Let's rewrite the linear regression model:

$$Y_i = \beta_0 + \beta_1 WBC_i + \varepsilon_i \quad E[Y_i | WBC_i] = \beta_0 + \beta_1 WBC_i$$

$$\varepsilon_i \stackrel{iid}{\sim} N(0, \sigma_\varepsilon^2)$$

$$\Rightarrow Y_i | WBC_i \sim N(\beta_0 + \beta_1 WBC_i, \sigma_\varepsilon^2)$$

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

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

Problem: $Y_i = 0 \text{ or } 1 \Rightarrow Y_i | WBC_i$ is not normal

Let's use Bernallli instead!

Second attempt

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

systematic component (wrong) $p_i = \beta_0 + \beta_1 WBC_i$

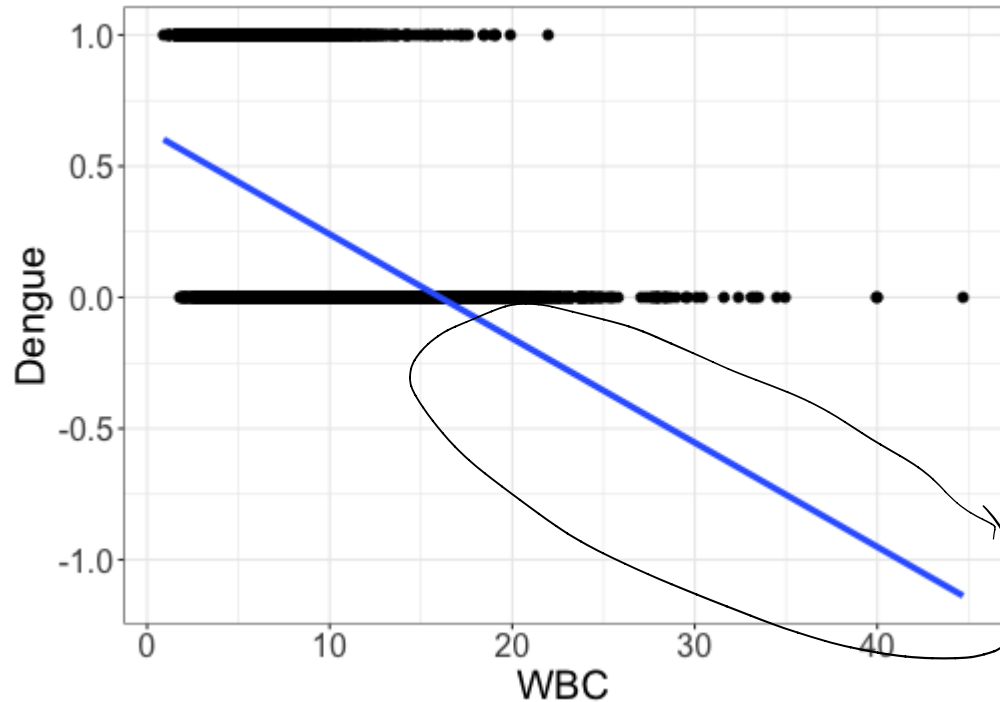
Are there still any potential issues with this approach?

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

$$\text{but } \beta_0 + \beta_1 WBC_i \in (-\infty, \infty)$$

(unless $\beta_1 \equiv 0$)

Don't fit linear regression with a binary response



IF $WBC > 15$,
predictions < 0

instead: fit a curve!

Fixing the issue: logistic regression

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

random component

$$g(p_i) = \beta_0 + \beta_1 WBC_i$$

systematic component

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

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

$$\frac{p_i}{1 - p_i} = \text{odds}$$

link function
links parameter p_i
to predictor WBC_i

log odds
aka logit

Odds

Definition: If $p_i = \mathbb{P}(Y_i = 1 | WBC_i)$, the **odds** are $\frac{p_i}{1 - p_i}$

Example: Suppose that $\mathbb{P}(Y_i = 1 | WBC_i) = 0.8$. What are the *odds* that the patient has dengue?

$$\text{odds} = \frac{0.8}{1 - 0.8} = \frac{0.8}{0.2} = 4$$

So, prob. patient has dengue = 4 × prob. patient does not have dengue

Odds

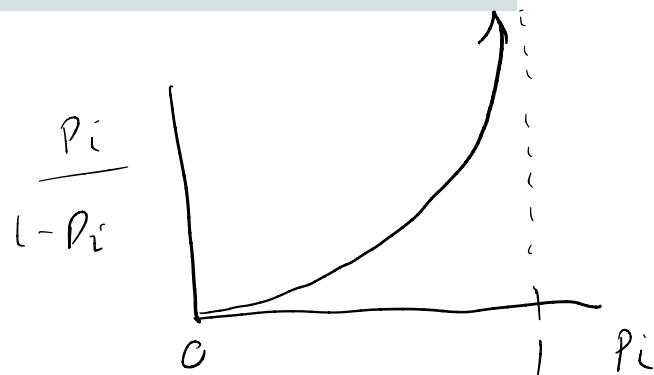
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The probabilities $p_i \in [0, 1]$. The linear function $\beta_0 + \beta_1 WBC_i \in (-\infty, \infty)$. What range of values can $\frac{p_i}{1 - p_i}$ take?

$$\text{if } p = 0 \quad \text{odds} = 0$$

$$\text{if } p = 1 \quad \text{odds} = \infty$$

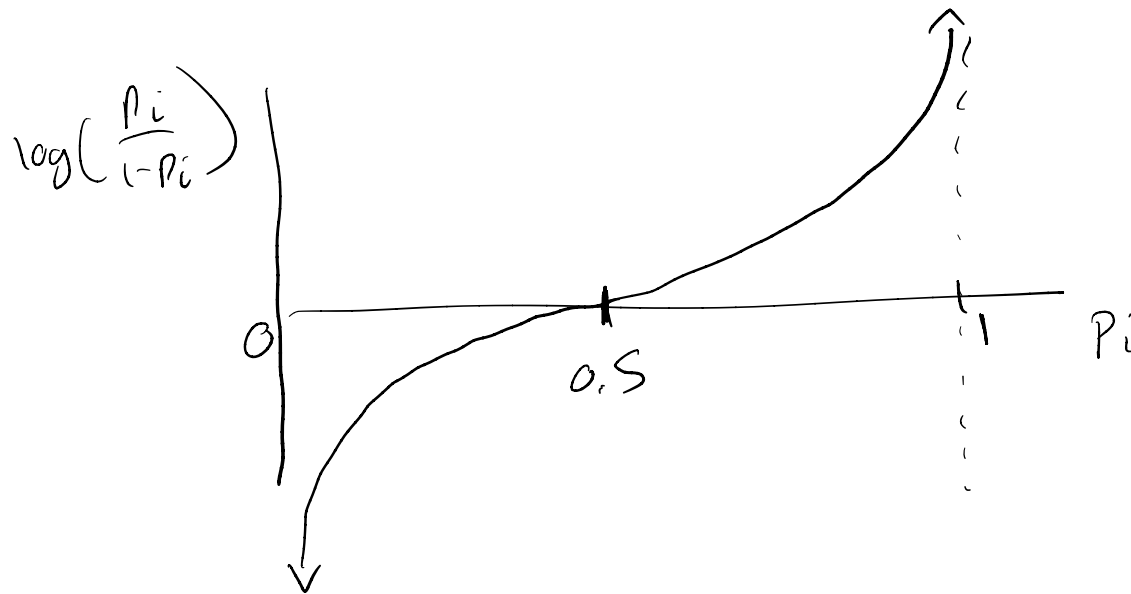
$$\text{odds} \in [0, \infty)$$



Log odds

$$g(p_i) = \log\left(\frac{p_i}{1-p_i}\right)$$

$$\frac{p_i}{1-p_i} \in [0, \infty) \Rightarrow \log\left(\frac{p_i}{1-p_i}\right) \in (-\infty, \infty) \quad \checkmark$$



Binary logistic regression

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

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

Note: Can generalize to $Y_i \sim \text{Binomial}(m_i, p_i)$, but we won't do that yet.

random component: specifies distribution of Y_i

systematic component: relates distribution to explanatory variables)

Example: simple logistic regression with dengue

Y_i = dengue status (0 = no, 1 = yes) $Y_i \sim \text{Bernoulli}(p_i)$

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

Work in groups of 2-3 for 5 minutes on the following questions:

- + Are patients with a higher WBC more or less likely to have dengue?
- + Interpret the estimated slope in context of a unit change in the log odds.
- + What is the change in *odds* associated with a unit increase in WBC?