

## CSC-591: Foundations of Data Science

T/Th. 12:50-2:05pm. EBI-1005.

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W6: 9/29/15-10/1/15

## Administrative

- Updated Weekly Schedule (on Moodle)
- 1<sup>st</sup> Midterm: 10/6/15
  - Review: 10/1/15 (important, don't miss)
- Additional Reading Materials
  - Logistic Regression, by Kleinbaum (Springer, Through NCSU Library)
  - Computing Primer for Applied Linear Regression, 4ed, Using R. <http://z.umn.edu/alrprimer>

## Regression, So far

- Linear Regression
  - Least Squares
- Correlation
- Regression Parameters
- Properties
- Significance of “r”
- Total Variation (explained + unexplained)
- Coefficient of Determination
- Standard Error of estimate, Prediction Interval
- Multiple Linear Regression
- Multiple Correlation Coefficient (R)
- Testing for significance of R

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## Regression As Classification

- So far, regression as prediction
- Today, regression as classification

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## Binary Classification Problems

- Problems that have two outcomes
- True/False categorical outcomes
  - Had Fever/no fever
  - Had a disease/no disease
  - (mechanical) failed/not failed
  - Win/loss
- Dichotomized categorical outcomes
  - Yes/no
  - Agree/Disagree

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## Recall That

- 0/1 Outcomes
  - Bernoulli outcomes
- Collection of exchangeable outcomes for same attribute (or covariate) data
  - Binomial outcome

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## Check if Regression Works

- Problem
  - Predicting NCSU Wolfpack (WF) game based on points (P)

$$WF_i = b_0 + b_1P_i + \epsilon_i$$

- $WF_i = \{0,1\}$



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## Observations

- Errors
  - Can't be normally distributed
- Error variance is not constant
  - It depends on the level of  $X_i$
- Response variable is bounded
  - $0 \leq Pr \leq 1$

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# Response Function

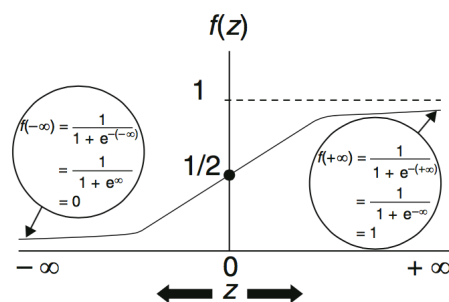
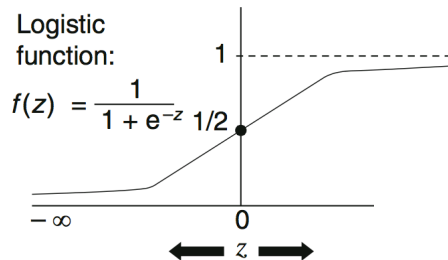
- Logit
- Probit
- Log-Log

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# Logistic Function



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Range:  $0 \leq f(z) \leq 1$

## Few Definitions

- Outcome (Binary; 0/1):  $WF_i(Y_i)$
- Probability (0, 1):  $\Pr(Y_i | X_i, b_0, b_1)$
- Odd (0,  $\infty$ ):  $(1/(1 - \Pr()))$ :  $\frac{\Pr(Y_i | X_i, b_0, b_1)}{1 - \Pr(Y_i | X_i, b_0, b_1)}$
- Log odds ( $-\infty, \infty$ ):  
(Logit)  $\log \frac{\Pr(Y_i | X_i, b_0, b_1)}{1 - \Pr(Y_i | X_i, b_0, b_1)}$

## Linear vs. Logistic Regression

- Linear:  $Y_i = b_0 + b_1 X_i + \varepsilon_i$   
$$E[Y_i | X_i, b_0, b_1] = b_0 + b_1 X_i$$
- Logistic:  $\Pr(Y_i | X_i, b_0, b_1) = \hat{\pi}_i = \frac{e^{(b_0 + b_1 X_i)}}{1 + e^{(b_0 + b_1 X_i)}}$

## Fitted Logit Response Function

$$\hat{\pi}_i = \frac{e^{(b_0 + b_1 X_i)}}{1 + e^{(b_0 + b_1 X_i)}}$$

- Log Odds:  $\hat{\pi}'_i = \log_e \frac{\hat{\pi}_i}{1 - \hat{\pi}_i}$

$$\hat{\pi}'_i = b_0 + b_1 X_i$$

## In Summary

- The Logistic model

$$z = \alpha + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_k X_k$$

$$f(z) = \frac{1}{1 + e^{-z}}$$

$$= \frac{1}{1 + e^{-(\alpha + \sum \beta_i X_i)}}$$

- In essence,  $z$  is an index that combines  $X$ s

## Epidemiology Example

- Let  $X_1, X_2, \dots, X_k$  are observations on a group of subject at Time  $T_0$
- For each of those observation, we also determined the disease status, as either 1 if “with disease” or 0 if “without disease”.
- Objective
  - We wish to use this information to describe the probability that the disease will develop during a defined study period, say  $T_0$  to  $T_1$ , in a disease- free individual with independent variable values  $X_1, X_2$ , up to  $X_k$ , which are measured at  $T_0$

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## Epidemiology Example

$$P(D = 1 | X_1, X_2, \dots, X_k)$$

$$= P(\mathbf{X})$$

Model formula:

$$P(\mathbf{X}) = \frac{1}{1 + e^{-(\alpha + \sum \beta_i X_i)}}$$

- Then, from observed data, we can estimate the parameters;  $b_0, b_1, b_2, \dots, b_k$
- For a given patient, we can estimate the risk by simply plug-in the observations ( $X_i$ ) into to model

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## Acknowledgements

- G. James, et. al., Moore, et. al.
- Kleinbaum, et. al.