

Logistic Regression

Data Science: Jordan Boyd-Graber

University of Maryland

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Data Science: Jordan Boyd-Graber | UMD Logistic Regression | 1 / 1

What are we talking about?

- Statistical classification: p(y|x)
- y is typically a Bernoulli or multinomial outcome
- Classification uses: ad placement, spam detection
- Building block of other machine learning methods

Logistic Regression: Definition

- Weight vector β_i
- Observations X_i
- "Bias" β_0 (like intercept in linear regression)

$$P(Y = 0|X) = \frac{1}{1 + \exp[\beta_0 + \sum_i \beta_i X_i]}$$
 (1)

$$P(Y=1|X) = \frac{\exp\left[\beta_0 + \sum_i \beta_i X_i\right]}{1 + \exp\left[\beta_0 + \sum_i \beta_i X_i\right]}$$
(2)

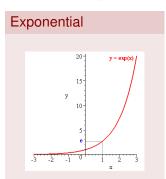
For shorthand, we'll say that

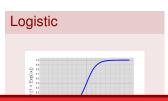
$$P(Y=0|X) = \sigma(-(\beta_0 + \sum_i \beta_i X_i))$$
(3)

$$P(Y = 1|X) = 1 - \sigma(-(\beta_0 + \sum_i \beta_i X_i))$$
 (4)

• Where $\sigma(z) = \frac{1}{1 + exp[-z]}$

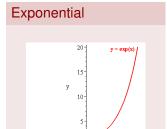
What's this "exp" doing?

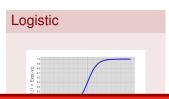




- $\exp[x]$ is shorthand for e^x
- e is a special number, about 2.71828
 - e^x is the limit of compound interest formula as compounds become infinitely small
 - It's the function whose derivative is itself.
- The "logistic" function is $\sigma(z) = \frac{1}{1+e^{-z}}$
- Looks like an "S"
- Always between 0 and 1.

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 - e^x is the limit of compound interest formula as compounds become infinitely small
 - It's the function whose derivative is itself.
- The "logistic" function is $\sigma(z) = \frac{1}{1+e^{-z}}$
- Looks like an "S"
- Always between 0 and 1.
 - Allows us to model probabilities
 - Different from linear regression

feature	coefficient	weight
bias	$oldsymbol{eta}_0$	0.1
"viagra"	$oldsymbol{eta}_1$	2.0
"mother"	eta_2	-1.0
"work"	eta_3	-0.5
"nigeria"	eta_4	3.0

■ What does Y = 1 mean?

Example 1: Empty Document? $X = \{\}$

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$$P(Y=0) = \frac{1}{1+\exp[0.1]} = 0.48$$

■
$$P(Y=1) = \frac{\exp[0.1]}{1 + \exp[0.1]} = 0.52$$

 Bias β₀ encodes the prior probability of a class

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Example 2 $X = \{Mother, Nigeria\}$

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Example 2

 $X = \{Mother, Nigeria\}$

•
$$P(Y=0) = \frac{1}{1+\exp[0.1-1.0+3.0]} =$$

•
$$P(Y=1) = \frac{\exp[0.1-1.0+3.0]}{1+\exp[0.1-1.0+3.0]} =$$

Include bias, and sum the other weights

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■ What does *Y* = 1 mean?

Example 2

 $X = \{Mother, Nigeria\}$

$$P(Y=0) = \frac{1}{1 + \exp[0.1 - 1.0 + 3.0]} = 0.11$$

$$P(Y=1) = \frac{\exp[0.1-1.0+3.0]}{1+\exp[0.1-1.0+3.0]} = 0.88$$

Include bias, and sum the other weights

feature	coefficient	weight
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Example 3 $X = \{Mother, Work, Viagra, Mother\}$

• What does Y = 1 mean?

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■ What does Y = 1 mean?

Example 3

 $X = \{Mother, Work, Viagra, Mother\}$

$$P(Y=0) = \frac{1}{1 + \exp[0.1 - 1.0 - 0.5 + 2.0 - 1.0]} =$$

$$\begin{array}{c} \bullet \ P(Y=1) = \\ \frac{\exp[0.1-1.0-0.5+2.0-1.0]}{1+\exp[0.1-1.0-0.5+2.0-1.0]} = \end{array}$$

 Multiply feature presence by weight

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■ What does Y = 1 mean?

Example 3

 $X = \{Mother, Work, Viagra, Mother\}$

$$P(Y=0) = \frac{1}{1 + \exp[0.1 - 1.0 - 0.5 + 2.0 - 1.0]} = 0.60$$

$$P(Y=1) = \frac{\exp[0.1-1.0-0.5+2.0-1.0]}{1+\exp[0.1-1.0-0.5+2.0-1.0]} = 0.30$$

 Multiply feature presence by weight