

Lesson Plan 4/12

Wednesday, April 11, 2018 12:09 PM

Admin

- HW4 due tonight
- HW5 posted
- Project 3 (first part) posted

?

HDSC:

- The 2nd generation p-value: <http://journals.plos.org/plosone/article?id=10.1371/journal.pone.0188299>

HW4

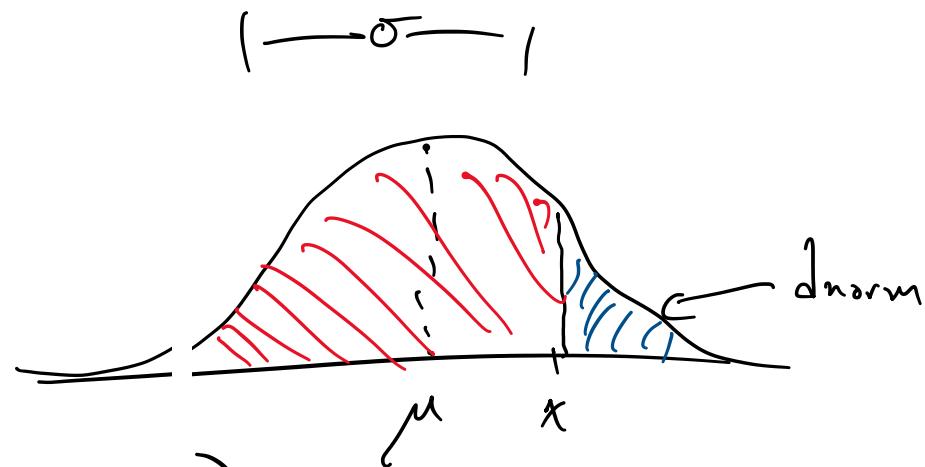
HW5

Project 3 (first part)

Logistic Regression

Solving linear ML problems

- p_{norm}
- d_{norm}
- g_{norm}



$$\text{pnorm}(x, \text{mean}=\mu, \text{sd}=\sigma) = \text{TRUE}$$

.7 ~ " " 1 - pnorm(x, mean=μ, sd=σ) = FALSE

probability distribution, "normal", "bell-shaped", μ

$$\text{red} + \text{blue} = 1 \quad P(\bar{X} > x) < .05$$

$$\int_{-\infty}^{\infty} d\text{norm} = 1 \quad \sim$$

density

$$d\text{norm}(x, \mu, \sigma) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left\{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2\right\}$$

f_{norm}

9

2

$$\text{lifeExp} = \beta_0 + \beta_1 \text{gdpPerCap}$$

↳ Classification

Logistic Regression

$$g = \begin{cases} 1 & \text{Yes} \\ 0 & \text{No} \end{cases} \Rightarrow y = \begin{cases} 1 & \text{if Yes} \\ 0 & \dots \end{cases}$$



Logistic regression

$$P(Y=1 | X_1=x_1, \dots, X_p=x_p)$$

$$\log \frac{P(Y=1 | X_1=x_1, \dots, X_p=x_p)}{P(Y=0 | X_1=x_1, \dots, X_p=x_p)} = \beta_0 + \beta_1 x_1 + \dots + \beta_p x_p$$

log odds

... it does not mean if

when

$$\log \frac{P(Y=1|x)}{P(Y=0|x)} = \underline{\log 1}$$

$$= 0$$

what does it mean; + -

$$\log \frac{P(Y=1|x)}{P(Y=0|x)} < 0 \Rightarrow \begin{matrix} \text{predict} \\ (\text{i.e., } Y=0) \end{matrix}$$

$$\log \frac{P(Y=1|x)}{P(Y=0|x)} > 0 \Rightarrow \begin{matrix} \text{predict Yes} \\ (\text{i.e., } Y=1) \end{matrix}$$

1 0 1 1

Linear Models

Regression $Y = \beta_0 + \beta_1 x_1 + \cdots + \beta_p x_p$

Class: $\log \frac{P(Y=1|X)}{P(Y=0|X)} = \beta_0 + \beta_1 x_1 + \cdots + \beta_p x_p$

Given $\{(x_1, y_1), \dots, (x_n, y_n)\}$,

Get estimates of $\hat{\beta}_0, \hat{\beta}_1, \dots, \hat{\beta}_p$

~~($\beta_0, \beta_1, \dots, \beta_p$)~~ / ~~Data~~ ^{Assumption}

$$RSS(\beta_0, \dots, \beta_p) = \sum_{i=1}^n (y_i - (\beta_0 + \beta_1 x_1 + \dots + \beta_p x_p))^2$$

↓
inverse

"loss function" (squared loss)

$$(\text{mean squared loss}) : \frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i)^2$$

Find parameters that minimize loss:

$$\arg \min_{\beta} \frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i(\beta))^2$$

Find values of β where "derivative" of loss is zero

Case study:

$$y = \beta x$$

Case