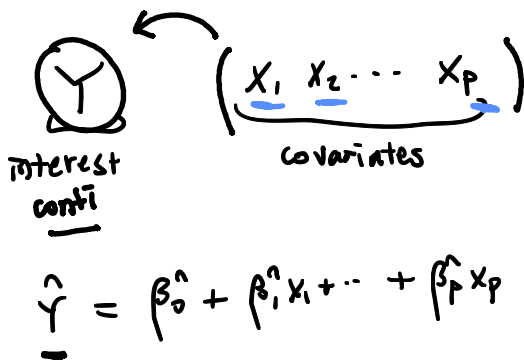


Week 12

$$\textcircled{Y} \rightarrow [\beta_0 + \beta_1 x_1 + \dots + \beta_p x_p]$$

Linear Model



Y: binary (0/1)

[dis. / no dis.
pass / fail]

$P(Y=1 | \cdot) = 0.7$

no accident : 0
accident : 1

$Y \sim x_1, x_2, x_3$

$\hat{P}(Y=1 | x_1, x_2, x_3)$

<u>Car Accident</u>	<u># of tickets</u>	<u>gender</u>	<u>age</u>
0	3	M	26
0	1	F	53
1	5	F	32
1	2	M	47
...
1	3	M	

new app

$\hat{P}(Y=1 | \cdot) = 0.9$

$\hat{P}(Y=1 | \cdot) = 0.4 - 1$

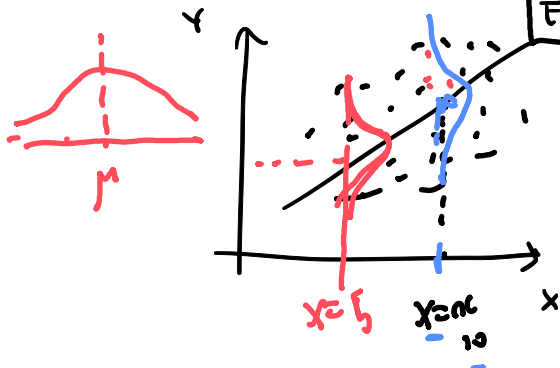
goal : Inference / prediction

Y and X
significant relationship?
(eg)
if yes, quantify

$\hat{P}(Y=1 | X=x)$

Parametric model
linear reg model

Y : conti



$$E(Y) = \beta_0 + \beta_1 X$$

$P(Y = \text{head}) = 0.2$
 $P(Y = \text{tail}) = 0.8$
HT

$Y | X = x \sim \text{Bernoulli}(P_{1x})$

(eg) coin toss

head $Y=1$ (Head)
tail $Y=0$
→ H H H T

specifically.

P_{1x}

$$Y | X = x \sim N(\beta_0 + \beta_1 x, \sigma^2)$$

Odds: $\frac{P}{1-P}$: (relative probability)

odds = 1
odds > 1
odds < 1

$$P(Y=1) = \frac{1}{2}$$

higher chance $Y=1$
 $P(Y=1) > \frac{1}{2}$

less chance $Y=1$
 $P(Y=1) < \frac{1}{2}$

$$\frac{P(\text{car accident})}{1 - P(\text{car accident})} = \frac{P(Y=1)}{1 - P(Y=1)}$$

$$Y = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$P(Y=1)$$

$$P(Y=0) + P(Y=1) = 1$$

$$\frac{P(Y=1)}{P(Y=0)} = \frac{P(Y=1)}{1 - P(Y=1)}$$

Odds Ratio

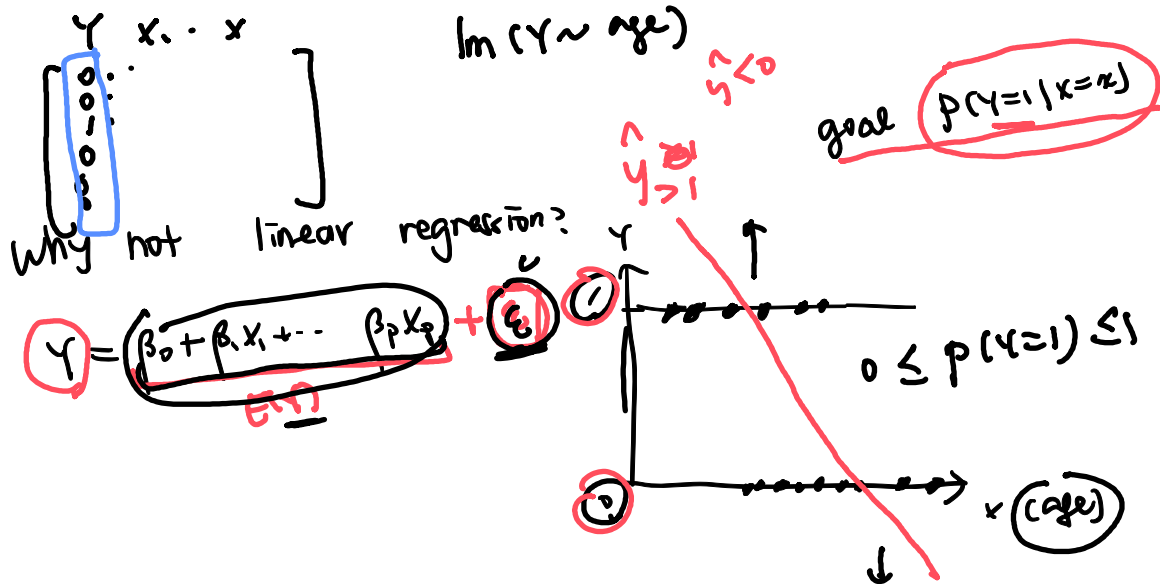
$$OR(\text{male vs. Female}) = \frac{\text{odds}(\text{car accident} | \text{Male})}{\text{odds}(\text{car accident} | \text{Female})} = 1$$

OR = 1
OR > 1
OR < 1

: no gender effect

Male > Female.

Male < Female



Logistic Regression
 $0 \leq P(Y=1) \leq 1$
 $\log\left(\frac{p}{1-p}\right)$
 $= (\beta_0 + \beta_1 x_1 + \dots + \beta_p x_p)$
 $-\infty < \log\left(\frac{p}{1-p}\right) < \infty$
 $-\infty < \beta_0 + \beta_1 x_1 + \dots + \beta_p x_p < \infty$
 $0 \leq p \leq 1$
 $0 \leq \frac{p}{1-p} < \infty$
 $0 \leq 1-p \leq 1$
 $-\infty < \log\left(\frac{p}{1-p}\right) < \infty$
 $p(Y=1 | x)$

Bernoulli $Y \sim \text{Bern}(0.8)$

prediction

$$\hat{p}(Y=1 | X=x)$$

$$\log\left(\frac{p}{1-p}\right) = \beta_0 + \beta_1 x_1 + \dots + \beta_p x_p$$

→ MLE

$$p(Y=1) = \frac{e^{\beta_0 + \beta_1 x_1 + \dots + \beta_p x_p}}{1 + e^{\beta_0 + \beta_1 x_1 + \dots + \beta_p x_p}}$$

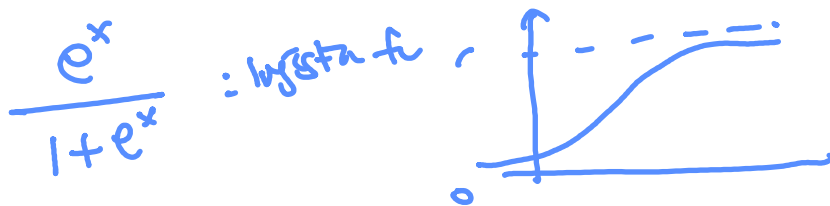
Given data, $\log\left(\frac{\hat{p}(Y=1|x)}{1-\hat{p}(Y=1|x)}\right) = \beta_0^{\wedge} + \beta_1^{\wedge} x_1 + \dots + \beta_p^{\wedge} x_p$

$$\frac{\hat{p}(Y=1|x)}{1-\hat{p}(Y=1|x)} = e^{\beta_0^{\wedge} + \beta_1^{\wedge} x_1 + \dots + \beta_p^{\wedge} x_p}$$

$$\hat{p}(Y=1|x) = \left(1 - \hat{p}(Y=1|x)\right) \cdot e^{\beta_0^{\wedge} + \beta_1^{\wedge} x_1 + \dots + \beta_p^{\wedge} x_p}$$

$$\hat{p}(Y=1|x) + \left(e^{\beta_0^{\wedge} + \beta_1^{\wedge} x_1 + \dots + \beta_p^{\wedge} x_p}\right) \hat{p}(Y=1|x) = e^{\beta_0^{\wedge} + \beta_1^{\wedge} x_1 + \dots + \beta_p^{\wedge} x_p}$$

$$\hat{p}(Y=1|x) = \frac{e^{\beta_0^{\wedge} + \beta_1^{\wedge} x_1 + \dots + \beta_p^{\wedge} x_p}}{1 + e^{\beta_0^{\wedge} + \beta_1^{\wedge} x_1 + \dots + \beta_p^{\wedge} x_p}}$$



Interpretation of β

$$\left[\log \left(\frac{p}{1-p} \right) = \beta_0 + \beta_1 x \right] \quad \beta_1$$

(i) (X_i) is categorical $\begin{cases} 0 : \text{male} \\ 1 : \text{female} \end{cases}$

odds $x=0$

$$\exp \log \left(\frac{p}{1-p} \mid \text{male} \right) = \beta_0$$

$$\text{odds}(\text{event} \mid \text{male}) = e^{\beta_0}$$

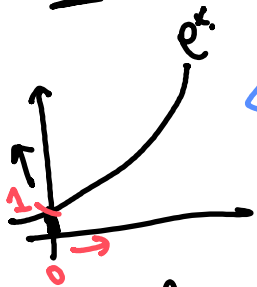
$x=1$

$$\exp \log \left(\frac{p}{1-p} \mid \text{female} \right) = \beta_0 + \beta_1$$

$$\text{odds}(\text{event} \mid \text{female}) = e^{\beta_0 + \beta_1}$$

$$\text{OR}(\text{female vs. male}) = \frac{\text{odds}(\text{event} \mid \text{female})}{\text{odds}(\text{event} \mid \text{male})}$$

$$e^{\beta_0} \cdot e^{\beta_1} = e^{\beta_0 + \beta_1} > 1$$



$$\beta_1 = 0 \Rightarrow e^{\beta_1} = 1$$

$$\beta_1 > 0 \Rightarrow e^{\beta_1} > 1$$

$$\beta_1 < 0 \Rightarrow e^{\beta_1} < 1$$

female > male

female < male

$$\text{odds}(\text{event} \mid \text{fem}) = \frac{e^{\beta_1} \cdot \text{odds}(\text{event} \mid \text{male})}{1}$$

↑ compare no gender

↑ ref

$e^0 = 1$

(ii) X IS CONTINUOUS

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

$x \rightarrow x+1$

age

$$X = x$$

exp $\log\left(\frac{p}{1-p}\right) = \beta_0 + \beta_1 x$
 $\text{odds}(x) = e^{\beta_0 + \beta_1 x}$

$$X = (x+1)$$

exp $\log\left(\frac{p}{1-p}\right) = \beta_0 + \beta_1 (x+1)$
 $\text{odds}(x+1) = e^{\beta_0 + \beta_1 (x+1)} = e^{\beta_0 + \beta_1 x + \beta_1}$

OR $(x+1 \text{ vs } x) = \frac{\text{odds}(x+1)}{\text{odds}(x)} = \frac{e^{\beta_0 + \beta_1 (x+1)}}{e^{\beta_0 + \beta_1 x}} = e^{\beta_1}$

$$\beta_1 = 0 \Rightarrow e^{\beta_1} = 1$$

no x effect
age

$$\left[\begin{array}{l} \beta_1 > 0 \Rightarrow e^{\beta_1} > 1 \\ \beta_1 < 0 \Rightarrow e^{\beta_1} < 1 \end{array} \right]$$

prob $x \uparrow$ | chance \uparrow
 $x \uparrow$ | \downarrow

$Y=1$ accident

$$\begin{cases} X_1 : \# \text{ of tickets} \\ X_2 : \text{gender (male}=0 / \text{Female}=1) \\ X_3 : \text{age} \end{cases} \quad \begin{cases} \hat{\beta}_1 = 2.3 \\ \hat{\beta}_2 = 1.5 \\ \hat{\beta}_3 = -1.3 \end{cases}$$

$$\checkmark \left[\text{OR}_{\text{ticket}} = \frac{\text{odds(event} | x+1)}{\text{odds(event} | x)} = e^{\hat{\beta}_1} = e^{2.3} = \underline{9.97} > 1 \right] \text{ more } \checkmark$$

$$\checkmark \left[\text{OR}_{\text{age}} = \frac{\text{odds(event} | x+1)}{\text{odds(event} | x)} = e^{\hat{\beta}_3} = e^{-1.3} = \underline{0.27} < 1 \right] \text{ younger}$$

$$\left[\text{OR}_{\text{gender}} = \frac{\text{odds(event} | \text{female})}{\text{odds(event} | \text{male})} = e^{1.5} = \underline{4.48} > 1 \right] \text{ F}$$

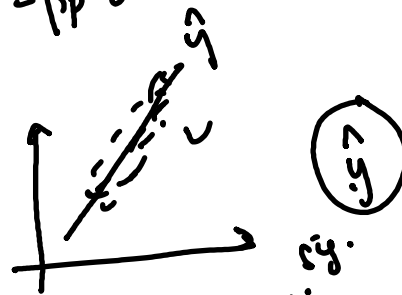
Who will have the highest chance of event?

Female. younger. ticket

Model Significance

- ✓ $\begin{cases} H_0: \text{model is not useful} \\ H_a: \text{model is useful} \end{cases}$

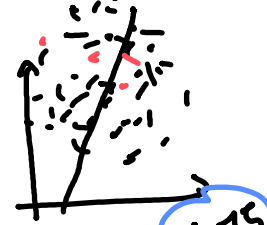
$$\beta_1 = \beta_2 = \dots = \beta_p = 0$$



→ Individual term significant

$H_0: \beta_f = 0$
 $H_a: \beta_f \neq 0$

$H_0: \beta_g = 0$
 $H_a: \beta_g \neq 0$



• Interpretation of β

β_{70}

$$OR_f = \frac{\text{odds}(\text{espr} = 1 | x = 1)}{\text{odds}(\text{espr} = 1 | x = 0)}$$

$$OR_g = \frac{odds(esr=1|x_H)}{odds(esr=1|x_L)}$$

$1.9 > 1$

0.155
71
11
1-16

Goodness-of-fit

pseudo R^2 SS ≈ 1

H_0 : model fits the data well
 H_a : " does "

$$\Rightarrow \hat{p}(Y=1|1.)$$

Model diagnostics

Assumption
{ data,
cook's

