

Topic 7

① `glm ()`.

② interpretation of coefficients in logistic model

③ ROC & AUC for logistic model

① glm ()

lm ()

glm (y ~ x₁ + x₂, data =)

Logistic model :

Response = categorical with 2 categories

→ should be numeric form : 0, 1

p denotes the probability response = 1.

$$\frac{p}{1-p}$$

odds of success

⇒ Logistic model :

$$\boxed{\log \frac{p}{1-p}} = \beta_0 + \beta_1 * \underline{x_1} + \beta_2 * \underline{x_2} \dots$$

churned ~ 1

② write down fitted model
& interpret coef :

\hat{p} = probability (churned = 1)

$$\underline{\underline{\log \frac{\hat{p}}{1-\hat{p}}}} = 3.4 - \underline{0.16} * \text{Age} + \underline{0.07} * \underline{1} (\text{Married} = 1) + \underline{0.02} * \text{years} + \underline{0.38} * \underline{\text{contacts}}$$

How to interpret 0.38 (coef of contacts) :

when no of contacts increases by 1 then the log-odds of churning increases by 0.38, keeping other variables the same.

$$\begin{aligned}\log(y_1) &= 5 + \textcircled{6}x \Rightarrow y_1 = e^{5+6x} \\ \log(y_2) &= 5 + 6(x+1) \Rightarrow y_2 = e^{5+6(x+1)} \\ &= 5 + 6x + \textcircled{6} \Rightarrow y_2 = e^{5+6x+6} \\ \frac{y_2}{y_1} &= e^{\textcircled{6}} \Rightarrow\end{aligned}$$

when the no of contacts \uparrow by 1, then the odds of churning changes by $e^{0.38}$ times, keeping other variables the same.

How to interpret 0.07 of Married?

Comparing a married person (Married = 1) vs a non-married when other variables are the same, then the log-odds of churning will be larger by 0.07.

Comparing between a married person vs non-married, then the odds of churning will change by $e^{0.07}$ times.

Column $\Pr(>|Z|)$ gives p-value for the test:

H_0 : variable is NOT significant | \rightarrow hope to get small
 H_1 : variable is significant | p-value.

③ ROC & AUC value for a logistic model.
 calculate $pr(Y = \text{yes})$

note: Model M_3 is built using train = full data set.
 → test the goodness of the model by considering test = full data set.

"prediction" from ROC:

Churned ←	prob ↓
0 ←	<u>0.08</u>
0 ←	0.1
1 ←	0.8
⋮	⋮

⋮

→ tpr, fpr.

Threshold

0.1

$prob(Y = \text{yes}) \geq 0.5 \rightarrow \text{Yes cate}$
 $\leq 0.5 \rightarrow \text{No cate}$

for Churn data: $\delta \approx 0.24$: classify as
 if predicted probability of churning $\geq \delta \rightarrow \text{Yes}$
 $< \delta \rightarrow \text{classifying as No.}$

8000

$\delta = 0.05 \rightarrow \text{tpr, fpr}$
 $\delta = 0.06 \rightarrow \text{tpr, fpr}$
 ⋮

7300 values of δ

