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title: "ISL_Chapter4_Logistic Regression"
author: "hyeju.kim"
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tags: [ISL]
image: LinearRegression.png
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<script type="text/javascript" src="http://cdn.mathjax.org/mathjax/latest/MathJax.js?config=TeX-AMS-MML_HTMLorMML"></script>

<script type="text/x-mathjax-config">

MathJax.Hub.Config({

    displayAlign: "center"

});

</script>


$$p(X) = \frac{e^{\beta_0 + \beta_1 X}}{1 + e^{\beta_0 + \beta_1 X}}$$


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Chapter 4. Classification

- What is classification?
predicting a qualitative response

4.1 An Overview of Classification

Dataset Introduction

- $\text{default}(Y)$: Yes / No
- $\text{balance}(X_1)$
- $\text{income}(X_2)$

4.2 Why Not Linear Regression?

$$Y = \begin{cases} 1 & \text{if stroke;} \\ 2 & \text{if drug overdose;} \\ 3 & \text{if epileptic seizure.} \end{cases}$$

- The gap between levels are not exactly same

Then, binary variable? (dummy variable)

- Estimates can be outside the $[0,1]$ interval

4.3 Logistic Regression

- Logistic regression model predicts **the probability that Y belongs to a particular category**, rather than the response Y directly

ex.

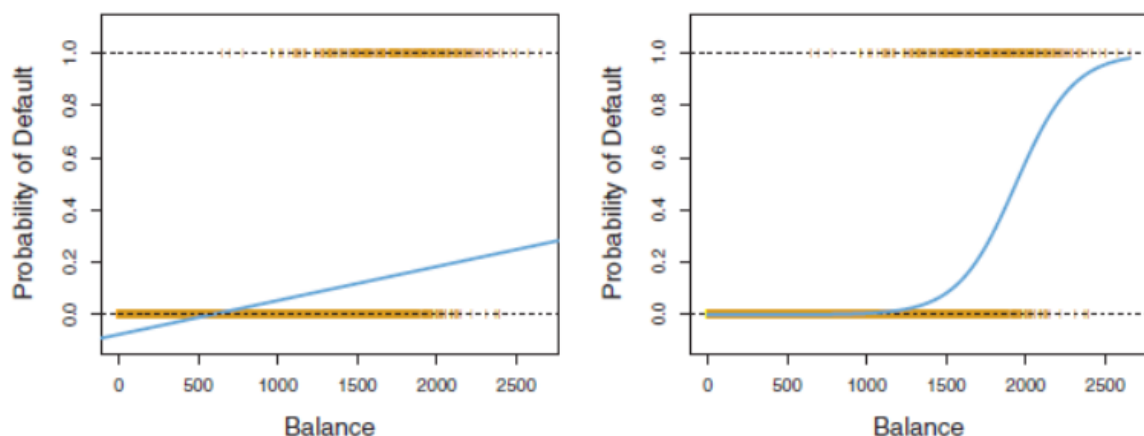


FIGURE 4.2. Classification using the **Default** data. Left: Estimated probability of **default** using linear regression. Some estimated probabilities are negative! The orange ticks indicate the 0/1 values coded for **default**(No or Yes). Right: Predicted probabilities of **default** using logistic regression. All probabilities lie between 0 and 1.

The Logistic Model

- how to set output values between [0,1]?

logistic function

$$p(X) = \frac{e^{\beta_0 + \beta_1 X}}{1 + e^{\beta_0 + \beta_1 X}} \quad (4.2)$$

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- S - shaped curve

After a bit of manipulation of (4.2), we find that

$$\frac{p(X)}{1 - p(X)} = e^{\beta_0 + \beta_1 X}. \quad (4.3)$$

- $p(X) / [1 - p(X)]$ is called **odds**, between 0 (very low possibility) and infinite (very high possibility)

By taking the logarithm of both sides of (4.3), we arrive at

$$\log \left(\frac{p(X)}{1 - p(X)} \right) = \beta_0 + \beta_1 X. \quad (4.4)$$

- $\log(p(X) / [1 - p(X)])$ is called the log-odds or **logit**.
- β_1 does not correspond to the change in $p(X)$ associated with a one-unit increase in X
- The amount that $p(X)$ changed due to one-unit change in X will depend on **the current value of X**
- **INCREASING X BY ONE UNIT CHANGES MULTIPLIES THE ODDS BY e^{β_1}**