

Modeling Binary Outcomes Using Logistic Regression with a Predictor

Part 3 - Estimating the Probability of Success

Learning Objectives

In this lecture, you will learn how to:

- Predict the probability of success using the logistic regression model
- Calculate its confidence interval

Example

A random sample of students is selected from a large statistics class.

The following variables are recorded:

- Number of hours studied
- Exam outcome, **pass (P)** or **fail (F)**

Hours	Grade
0	F
0	F
0.5	F
1.5	F
1.5	F
1.5	P
2	F
2.5	F
2.5	F
⋮	⋮
10.5	P
11	P
11	P

The full dataset 'Hours-and-Grades' can be downloaded from Brightspace

Modeling

The objective of using the data is to use the **number of hours** as a **predictor** to model the **probability of passing**.

The model described is a **Logistic Regression model**, which relates the **log-odds** of **passing** to a **linear function of study hours**.

$$\underbrace{\ln \left(\frac{p}{1-p} \right)}_{\text{log-odds function of } p} = \underbrace{A + B * \textit{Hours}}_{\text{the linear function of the predictor (e.g., hours) as used in a standard regression model}}$$

Fitting a Logistic Regression Model to Data using R

We fit the logistic regression model to the data and obtain the following output.

```
Call:
glm(formula = y ~ x, family = binomial)

Coefficients:
              Estimate Std. Error z value Pr(>|z|)
(Intercept)  -2.8984      0.9694  -2.990 0.002791 **
x              0.6734      0.1860   3.621 0.000294 ***
---

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 64.104  on 49  degrees of freedom
Residual deviance: 36.354  on 48  degrees of freedom
AIC: 40.354
```

- Once we have determined that the **number of hours** is significant predictor of a student's **probability of passing the exam**,,
- we can use the **regression equation** to predict this **probability** based on the **number of hours studied**.
- To do that, we read the **coefficient table** from the regression output and use the values in the "Estimate" column

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	-2.8984	0.9694	-2.990	0.002791	**
x	0.6734	0.1860	3.621	0.000294	***

A (Intercept) = -2.8984, B (Slope) = 0.6734



$$\ln\left(\frac{p}{1-p}\right) = -2.8984 + 0.6734 * \text{Hours}$$

Let's re-write the regression equation in terms of p

$$\ln \left(\frac{p}{1-p} \right) = -2.8984 + 0.6734 * \textit{Hours}$$



$$p = \frac{e^{-2.8984+0.6734*\textit{Hours}}}{1 + e^{-2.8984+0.6734*\textit{Hours}}}$$

Now, we want to estimate the **probability** that a student will pass the exam if the student studies for **0 hours**.

$$\begin{aligned} p &= \frac{e^{-2.8984 + \overbrace{0.6734 \cdot \text{Hours}}^0}}{1 + e^{-2.8984 + \underbrace{0.6734 \cdot \text{Hours}}_0}} \\ &= \frac{e^{-2.8984}}{1 + e^{-2.8984}} \\ &\approx 0.055 \end{aligned}$$

If a student studies for **0 hours**, the estimated **probability of passing the exam** is 0.055.

Confidence Interval for the Probability

- The **estimated probability** we calculated always differs from its **true value**.
- Therefore, we always use a **confidence interval** that provides a range of possible value for the **probability**.
- The **confidence interval** for the true **probability** at a given x-value, x_0 is calculated in four steps.

Step 1 - Calculate the estimated log-odds from the regression equation

$$\text{Estimated } \underbrace{\ln \left(\frac{p}{1-p} \right)}_{\text{log-odds}} = \hat{A} + \hat{B} x_0$$

Confidence Interval for the Probability

Step 2 - Calculate the variance of the estimated log-odds

$$\begin{aligned} \text{Var} \left[\underbrace{\text{Estimated} \ln \left(\frac{p}{1-p} \right)}_{\text{log-odds}} \right] &= \text{Var}[\hat{A} + \hat{B} x_0] \\ &= \text{Var}[\hat{A}] + \text{Var}[\hat{B}](x_0)^2 + 2x_0 \text{COV}(\hat{A}, \hat{B}) \end{aligned}$$

OR written in a **Matrix form** if you are good in **Matrix algebra**

$$= [1, x_0] \underbrace{\begin{bmatrix} \text{Var}[\hat{A}] & \text{COV}(\hat{A}, \hat{B}) \\ \text{COV}(\hat{A}, \hat{B}) & \text{Var}[\hat{B}] \end{bmatrix}}_{\text{Covariance Matrix}} \begin{bmatrix} 1 \\ x_0 \end{bmatrix}$$

We will use R to obtain the variance of \hat{A} / \hat{B} and covariance of \hat{A} and \hat{B}

There is no need to do any hand calculation.

Confidence Interval for the Probability

Step 3 - Get the Confidence Interval for log-odds

The Confidence Interval for **log-odds** is given by:

$$\underbrace{\text{Estimated } \ln\left(\frac{p}{1-p}\right)}_{\text{log-odds}} \pm z_c \times \sqrt{\text{Var}\left[\underbrace{\text{Estimated } \ln\left(\frac{p}{1-p}\right)}_{\text{log-odds}}\right]}$$



$$= \left(\hat{A} + \hat{B} x_0 \right) \pm z_c \times \sqrt{\text{Var}[\hat{A}] + \text{Var}[\hat{B}](x_0)^2 + 2x_0 \text{COV}(\hat{A}, \hat{B})}$$

Confidence Interval for the Prediction

Last Step

- Let (L, U) represent the lower and upper bounds of the confidence interval for the *log-odds* calculated in the previous step.
- Recall that the probability of success is given by:

$$p = \frac{e^{A+Bx}}{1 + e^{A+Bx}} = \frac{e^{\text{log-odds}}}{1 + e^{\text{log-odds}}}$$

The confidence interval for the true odds is given by:

$$\left(\frac{e^L}{1 + e^L}, \quad \frac{e^U}{1 + e^U} \right)$$

Confidence Interval for the Prediction

From the previous analysis, we estimated that the **probability of passing** is **0.055** if a student studies for **0 hours**.

Question: Construct a 95% confidence interval for the **true probability**

Step 1 - Calculate the estimated log-odds from the regression equation

$$\ln\left(\frac{p}{1-p}\right) = -2.8984 + 0.6734 * \underbrace{\text{Hours}}_0 \rightarrow$$
$$= -2.8984$$

Confidence Interval for the Prediction

Step 2 - Calculate the variance of the estimated log-odds

First, we need to use R to get the variance of \hat{A} / \hat{B} and the covariance of \hat{A} and \hat{B} .

The following is the R command:

```
cov.matrix = vcov( fitted.model )
```



R Console

```
> cov.matrix = vcov( fitted.model)
> cov.matrix
```

	(Intercept)	x
(Intercept)	0.9397409	-0.16273699
x	-0.1627370	0.03459715

```
> cov.matrix = vcov( fitted.model)
> cov.matrix
              (Intercept)              x
(Intercept)  0.9397409 -0.16273699
x            -0.1627370  0.03459715
```

In the matrix,

- $\text{Var}(\hat{A}) = \text{Var}(\text{intercept}) = 0.93974$
- $\text{Var}(\hat{B}) = \text{Var}(\text{slope}) = 0.34597$
- $\text{Cov}(\hat{A}, \hat{B}) = -0.16274$

$$\begin{aligned}
 \text{So, } \text{Var}\left[\underbrace{\text{Estimated } \ln\left(\frac{p}{1-p}\right)}_{\text{log-odds}}\right] &= \text{Var}[\hat{A} + \hat{B} x_0] \\
 &= \underbrace{\text{Var}[\hat{A}]}_{0.93974} + \underbrace{\text{Var}[\hat{B}]}_{0.34597} \underbrace{(x_0)^2}_{0 \text{ hours}} + 2x_0 \underbrace{\text{COV}(\hat{A}, \hat{B})}_{-0.16274} \\
 &= 0.93974
 \end{aligned}$$

Note: You can use R to perform matrix calculation of variance.

$$\text{Var} \left[\text{Estimated} \ln \left(\frac{p}{1-p} \right) \right]_{\text{log-odds}} = [1, x_0] \begin{bmatrix} \text{Var}[\hat{A}] & \text{COV}(\hat{A}, \hat{B}) \\ \text{COV}(\hat{A}, \hat{B}) & \text{Var}[\hat{B}] \end{bmatrix} \begin{bmatrix} 1 \\ x_0 \end{bmatrix}$$

The following is the command:

```
x0 = 0
xvector = matrix( c(1, x0) )
var.log.odds = t(xvector) %*% cov.matrix %*% xvector
var.log.odds
```

R Console

```
> x0 = 0
> xvector = matrix( c(1, x0) )
> var.log.odds = t(xvector) %*% cov.matrix %*% xvector
> var.log.odds
      [,1]
[1,] 0.9397409
```


Step 3 - Get the Confidence Interval for log-odds

$$\begin{aligned} & \underbrace{\text{Estimated } \ln\left(\frac{p}{1-p}\right)}_{\text{log-odds}} \pm z_c \times \sqrt{\text{Var}\left[\underbrace{\text{Estimated } \ln\left(\frac{p}{1-p}\right)}_{\text{log-odds}}\right]} \\ & \quad \downarrow \qquad \qquad \downarrow \qquad \qquad \downarrow \\ & = -2.8984 \pm 1.96 \times \sqrt{0.93974} \\ & = (-4.7984, -0.9983) \end{aligned}$$

Last Step - Get the Confidence Interval for odds

$$\left(\frac{e^{-4.7984}}{1 + e^{-4.7984}}, \frac{e^{-0.9983}}{1 + e^{-0.9983}} \right) = (0.0082, \quad 0.2693)$$

If a student studies for **0 hours**,

We are 95% confident that there is between **0.0082 and 0.2693**

probability that **this student will pass the exam**

Confidence Interval for the Prediction

Exercise: Suppose a student studies for 1 hour.

a) Estimate the probability that the student will pass the exam.

Please perform the calculation manually (by hand)

b) Use R to construct a 95% confidence interval for the true probability of a student passing the exam