

# Assumptions

Wednesday 1 May 2024 9:34 PM

**NOTE:** violating these assumptions doesn't mean we can't use logistic regression, but it may impact the validity of the results & we should proceed with caution.

1) Binary logistic regression requires the dependent variable to be binary  
i.e.  $y \rightarrow$  bernoulli distribution

2) Independence of observations

3) Absence of multicollinearity in independent variables

4) large sample size

5) linearity of independent variables and log odds

→ Odds: the odds of an event is the ratio of the probability of the event happening to the probability of the event not happening. It's a way of expressing the likelihood of an event. If the odds are greater than 1, the event is more likely to happen than not.

$$\left| \frac{P}{1-P} \right| [0 \rightarrow \infty)$$

$\log(\text{odds}) \rightarrow$  brings symmetry to scale

log(odds) connection to logistic regression?

$$\log(\text{odds}) = \log\left(\frac{P}{1-P}\right)$$

$x_1$	$x_2$	$x_3$	$x_4$	$\hat{y}$

$$\hat{y} = P(1) = P = \frac{1}{1 + e^{-\beta x}}$$

$$P = \frac{1}{1 + e^{-\beta x}}$$

$$1 + e^{-\beta x} = \frac{1}{P}$$

$$e^{-\beta x} = \frac{1}{P} - 1$$

$$\rightarrow e^{-\beta x} = \frac{1-P}{P}$$

$$\frac{1}{e^{\beta x}} = \frac{1-P}{P}$$

$$\frac{P}{1-P} = e^{\beta x}$$

$$\log\left(\frac{P}{1-P}\right) = \beta x$$

$$\log(\text{odds}) = \beta x$$

$$\left| P = \frac{1}{1 + e^{-\log(\text{odds})}} \right|$$

