Article

Mathematical methodology and *MATLAB* computer program to calculate the effective-dose of percent response using the probit analysis technique

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Abstract

Using the probit analysis technique, mathematical algorithm and *MATLAB* computer program have been implemented in this paper to calculate both the Log-Dose (*LD*) and the Effective-Dose (*ED*) for any given percent. The probit analysis uses a successive weighted simple linear regression of experimental or row data. The kind of row data obtained from the bioassays should be generally in percent response (mortality or affected) at the corresponding doses (or concentrations). The response should be always in binomial form (e.g. death/no death) and the relationship between the response and the various doses or concentrations is always sigmoid or non-linear. The probit analysis here acts as a transformation from sigmoid or non-linear relationship to linear one and then uses a successive weighted simple linear regression on the linear relationship between number responding (not proportion response) and dose (or concentration) should be normally distributed. Two simple examples are explained in this paper to prove the validity and the consistency of both the proposed mathematical methodology and the concerning computer program.

Keywords probit analysis; weighted simple linear regression; bioassay; transformation; Log-Dose; Effective-Dose; Log-Concentration; Effective-Concentration.

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1 Introduction

The probit analysis technique is considered as the most popular approach and methodology to modeling binomial response variables. It used extensively in the biological sciences, biomedical research, toxicology, and engineering. Indeed, even in the social sciences binomial response variables are found to be plentiful. The basic distribution for the response is either Bernoulli or binomial. The probit analysis transforms the sigmoid

dose-response curve to a straight line that can then be analyzed by regression either through least squares method or maximum likelihood function.

The idea of probit analysis was originally published in Science by Chester Ittner Bliss, 1934. He worked as an entomologist for the Connecticut agricultural experiment station and was primarily concerned with finding an effective pesticide to control insects that fed on grape leaves (Greenberg, 1980). By plotting the response of the insects to various concentrations of pesticides, he could visually see that each pesticide affected the insects at different concentrations, i.e. one was more effective than the other. However, he didn't have a statistically sound method to compare this difference. The most logical approach would be to fit a regression of the response versus the concentration, or dose and compare between the different pesticides. Yet, the relationship of response to dose was sigmoid or non-linear in nature and at the time regression was only used on linear data. Therefore, Bliss developed the idea of transforming the sigmoid dose-response curve to a straight line. In 1952, a professor of statistics at the University of Edinburgh by the name of David Finney took Bliss' idea and wrote books called Probit Analysis (Finney, 1952, 1964). Today, probit analysis is still the preferred statistical method in understanding dose-response relationships.

Probit analysis is also commonly used in toxicology to determine the relative toxicity of chemicals to living organisms. This is done by testing the response of an organism under various concentrations of each of the chemicals in question and then comparing the concentrations at which one encounters a response. The response is always binomial (e.g. death/no death) and the relationship between the response and the various concentrations is always sigmoid or non-linear. Probit analysis acts as a transformation from sigmoid to linear and then runs a regression on the relationship.

Once a regression is run, the researcher can use the output of the probit analysis to compare the amount of chemical required to create the same response in each of the various chemicals. There are many endpoints used to compare the differing toxicities of chemicals, but the Log-Concentration (*percent*) or in abbreviation *LC* (*percent*) or Log-Dose (*percent*) and in abbreviation *LD* (*percent*) are the most widely used outcomes of the dose-response experiments. The *LC* (*percent*)/*LD* (*percent*) represent the concentration (*LC* (*percent*)) or dose (*LD* (*percent*)) at which percent of the population responds.

After knowing the *LC* (*percent*)/*LD* (*percent*), it easily to be calculated the effective- Concentration (*percent*) and in abbreviation *EC* (*percent*) and the Effective-Dose (*percent*) and in abbreviation *ED* (*percent*).

For example, consider comparing the toxicity of two different pesticides to aphids, pesticide *A* and pesticide *B*. If the *LC* (0.50) of pesticide *A* is 50 µg/L and the *LC* (0.50) of pesticide *B* is 10 µg/L, pesticide *B* is more toxic than *A* because it only takes 10ug/L to kill 50% of the aphids, versus 50 µg/L of pesticide *B*. The same explanation for the *EC* (0.50) which is easily results from *LC* (0.50).

For more details, profound and expanded explanations about the probit analysis, readers are invited to consult the papers Throne et al. (1995), Moermans and Nelis (1994), Sakuma (1998), Kumar et al. (2020) and Vincent (2008).

2 Methodology and Algorithm

Input Data: Input the vectors: concentration, proportion, total number

Imput the numbers: total iterations, percent

where

k is the number of iterations *m* is the number of groups total number (*i*) is the total number in the group (*i*) (*i*=1,...,*m*) percent is 50%, 60%, ε is a small positive number close to zero

Replacement:Set c(i)=concentration(i) for i=1,...,mSet p(i)=proportion(i) for i=1,...,mSet n(i)=total number(i) for i=1,...,m

Initial iteration (k = 1)

Calculate the vector:
$$x(i) = \log_2\left(\frac{c(i)}{c(1)}\right)$$
 $(i = 1,...,m)$

Calculate the vector: $y(i) = \operatorname{norminv}[p(i)](i = 1, ..., m)$

Calculate the vector:
$$z(i) = \frac{1}{\sqrt{2\pi}}e^{-\frac{(y(i))^2}{2}}$$
 $(i = 1, ..., m)$

Calculate the vector: $w(i) = \frac{n(i)(z(i))^2}{p(i)(1-p(i))}$ (i = 1,...,m)

Solve the following set of linear equations:

$$\begin{bmatrix} \sum_{i=1}^{m} w(i) & \sum_{i=1}^{m} w(i)x(i) \\ \sum_{i=1}^{m} w(i)x(i) & \sum_{i=1}^{m} w(i)(x(i))^{2} \end{bmatrix} \begin{bmatrix} \alpha \\ \beta \end{bmatrix} = \begin{bmatrix} \sum_{i=1}^{m} w(i)y(i) \\ \sum_{i=1}^{m} w(i)x(i)y(i) \end{bmatrix}$$

To find the intercept $\alpha^1 = \alpha$ and the slope $\beta^1 = \beta$

Put k = 1Iteration (k) While ($k \le totaliterations$) do

Calculate the vector: $Y^{k}(i) = \alpha^{k} + \beta^{k} x(i)$ (i = 1,...,m)

Calculate the vector: $P^{k}(i) = normcdf[Y^{k}(i)]$ (i = 1,...,m)

Calculate the vector: $Z^{k}(i) = \frac{1}{\sqrt{2\pi}}e^{-\frac{(Y^{k}(i))^{2}}{2}}$ (i = 1, ..., m)

Calculate the vector:
$$W^{k}(i) = \frac{n(i)(Z^{k}(i))^{2}}{P^{k}(i)(1-P^{k}(i))}$$
 $(i = 1,...,m)$

Solve the coming set of linear equations:

$$\begin{bmatrix} \sum_{i=1}^{m} W^{k}(i) & \sum_{i=1}^{m} W^{k}(i) x(i) \\ \sum_{i=1}^{m} W^{k}(i) x(i) & \sum_{i=1}^{m} W^{k}(i) (x(i))^{2} \end{bmatrix} \begin{bmatrix} \alpha \\ \beta \end{bmatrix} = \begin{bmatrix} \sum_{i=1}^{m} W^{k}(i) y(i) \\ \sum_{i=1}^{m} W^{k}(i) x(i) y(i) \end{bmatrix}$$

To find the intercept $\alpha^{k+1} = \alpha$ and the slope $\beta^{k+1} = \beta$

Stopping criteria

If (k = totaliterations) then stop or,

If
$$\left(\sqrt{\left(\alpha^{k+1} - \alpha^{k}\right)^{2} + \left(\beta^{k+1} - \beta^{k}\right)^{2}} < \varepsilon\right)$$
 then stop
Else put $k = k + 1$

End while

Set intercept= α^k and slop= β^k

Calculate the Log-Dose by: $LD(percent) = \frac{\text{norminv}(percent) - \text{intercept}}{2}$ slope

Calculate the effective-Dose by: $ED(percent)=c(1)\times 2^{LD(percent)}$

Calculate the variance of the Log-Dose by:

Variance[LD(percent)] =
$$\frac{1}{(\beta^{k})^{2}} \left[\frac{1}{\sum_{i=1}^{m} W^{k}(i)} + \frac{\left[\overline{x} - LD(percent)\right]^{2}}{\sum_{i=1}^{m} W^{k}(i) \left[x(i) - \overline{x}\right]^{2}} \right]$$

Where:
$$\overline{x} = \frac{\sum_{i=1}^{m} W^{k}(i) x(i)}{\sum_{i=1}^{m} W^{k}(i)}$$
 and

$$\sum_{i=1}^{m} W^{k}(i) [x(i) - \overline{x}]^{2} = \sum_{i=1}^{m} W^{k}(i) [x(i)]^{2} - \frac{\left[\sum_{i=1}^{m} W^{k}(i) x(i)\right]^{2}}{\sum_{i=1}^{m} W^{k}(i)}$$

Calculate the standard deviation of LD(percent) by:

$$\sigma[LD(percent)] = \sqrt{Variance[LD(percent)]}$$

Calculate the standard deviation of the concentration or dose by:

$$\sigma[ED(percent)] = ED(percent) \times \sigma[LD(percent)]$$

3 Notes and General Remarks

- 1. Impute the dose or the concentration vector in increasing order from small to large concentration(dose) including the control concentration(dose)
- 2. If there is a zero in the vectors: concentration or proportion, it should be replaced by ε where ε is a small positive number close to zero to respect the logarithm function

- 3. If there is a one (one hundred percent) in the vectors: concentration and proportion, it should be replaced by $(1-\varepsilon)$ where ε is a small positive number close to zero to respect the logarithm function
- 4. There is not any changes if we replace $\log_2 by \log_e or \log_{10} in$ the methodology, we can get the same results
- 5. We can use another weighting factors as: $W(i) = \frac{1}{\sqrt{n(i)P(i)[1-P(i)]}}$ (i = 1,...,m) or any

other mathematical formulas in the weighting simple linear regression

6. The reserved function "normcdf" in *MATLAB* is the normal cumulative distribution function and it is defined by:

$$p(i) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{y(i)} e^{-\frac{t^2}{2}} dt = normcdf [y(i)] \quad (i = 1, ..., m)$$

7. The reserved function "norminv" in *MATLAB* is the inverse of the normal cumulative distribution function and it is defined by:

$$prob[-\infty < Z < y(i)] = p(i) \equiv y(i) = norminv[p(i)]$$
 $(i = 1,...,m)$

where Z the standard normal variable with an arithmetic mean is zero and a standard deviation is one

4 Testing Validity and Accuracy of Both Methodology and Computer Code

Two examples are treated in this work to prove the validity and accuracy of the methodology and the concerning computer code (see supplementary material).

Example 1

The data set in Table 1 below shows the effect different doses of nicotine on the common fruit fly. The purpose of the experiment was to use the probit analysis to arrive at an appropriate model relating probability of "kill" to concentration. The analyst sought the *LD* (*percent*), that is, the concentration of nicotine that results in a certain probability. Of particular interest is the *LD* (*percent*), the concentration that produces a percent probability of "insect kill".

| | | _ | |
|---------------------|-------------------|--------------------------|-----------------------|
| Concentration(dose) | Number of insects | Number of insects killed | Proportion of insects |
| level | In group | In group | killed in group |
| 0.10 | 47 | 8 | 0.170 |
| 0.15 | 53 | 14 | 0.264 |
| 0.20 | 55 | 24 | 0.436 |
| 0.30 | 52 | 32 | 0.615 |
| 0.50 | 46 | 38 | 0.826 |
| 0.70 | 54 | 50 | 0.926 |
| 0.95 | 52 | 50 | 0.962 |

Table 1 Data set of example 1.

Imput data of the program are:

concentration=[0.1 0.15 0.20 0.30 0.50 0.7 0.95] "concentration or dose vector" proportion=[0.170 0.264 0.436 0.615 0.826 0.926 0.962] total number=[47 53 55 52 46 54 52]"numbers in groups" total iterations=100 percent=0.50 and percent=0.70

Results produced by the program are:

LD(0.50)=1.2143 with $\sigma[LD(0.50)]=0.0918$

ED(0.50)=0.2320 with $\sigma[ED(0.50)]=0.0213$ Versus the effective dose value 0.23176

reported by the SPSS statistical package

LD(0.70)=1.8183 with $\sigma[LD(0.70)]=0.0982$

ED(0.70)=0.3527 with $\sigma[ED(0.70)]=0.0346$ Versus the effective dose value 0.35273 reported by

the SPSS statistical package

Example 2

A typical set of data with groups' number of 20 larvae tested for each dose or concentration is given in Table 2 below.

| Concentration(dose) | Number of larvae | Number of larvae killed | Proportion of larvae |
|---------------------|------------------|-------------------------|----------------------|
| level | In group | In group | killed |
| | | | In group |
| 0.00375 | 20 | 0 | 0 |
| 0.0075 | 20 | 1 | 0.05 |
| 0.015 | 20 | 8 | 0.40 |
| 0.03 | 20 | 11 | 0.55 |
| 0.06 | 20 | 16 | 0.80 |
| 0.12 | 20 | 18 | 0.90 |
| 0.24 | 20 | 20 | 1.00 |

 Table 2 Data set of example 2.

Imput data of the program are:

concentration=[0.00375 0.0075 0.015 0.03 0.06 0.12 0.24] "concentration or dose vector" proportion=[0.001 0.05 0.40 0.55 0.80 0.90 0.999] total number=[20 20 20 20 20 20 20]"numbers in groups" total iterations=100

percent=0.50 and percent=0.80

Results produced by the program are:

LD(0.50)=2.8391 with $\sigma[LD(0.50)]=0.1885$

ED(0.50)=0.0268 with $\sigma[ED(0.50)]=0.0051$ Versus the effective dose value 0.0265 reported by

the SPSS statistical package

LD(0.80)=3.9090 with $\sigma[LD(0.80)]=0.2391$

ED(0.80)=0.0563 with $\sigma[ED(0.80)]=0.0135$ Versus the effective dose value 0.05913 reported

by the SPSS statistical package

5 Conclusions

An easy mathematical methodology and interactive computer software to calculate both the Log-Dose (*LD*) and the Effective-Dose (*ED*) for any given percent have been proposed in this paper. The proposed methodology and the code algorithm are mainly based on famous mathematical methodology well-known in applied statistics as probit analysis. The probit analysis acts as a transformation from sigmoid or non-linear relationship to linear one and then uses a successive weighted simple linear regression on the linear relationship of experimental or row data.

Two simple examples are shown to prove the validity and accuracy of both the proposed methodology and the concerned computer code. The code is implemented and programmed in *MATLAB* environment; it is easy to use and easy to be developed. The software is appended and it can be loaded directly when it is needed.

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