Machine Learning with R: Logistic Regression

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March 28: Intermediate Python Programming – Seyed Amirreza Mousavi

April 30: Survival Analysis with R – Humayun Kabir





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- ☐ Creating data visualizations, including charts, graphs, and scatter plots
- □ Figuring out which statistical tests to run (e.g., t-test, chi-square, etc.).
- $\ oxdot$ Analyzing data with software including SPSS, Python, R, SAS, ArcGIS, MATLAB, and Excel
- Choosing which software package to use, including free and open-source software
- ☐ Troubleshooting problems related to file formats, data retrieval, and download
- ☐ Selecting methodology and type of data analysis to use in a thesis project

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Machine Learning with R: Logistic Regression





Objective





LEARNING THE BASICS OF LOGISTIC REGRESSION

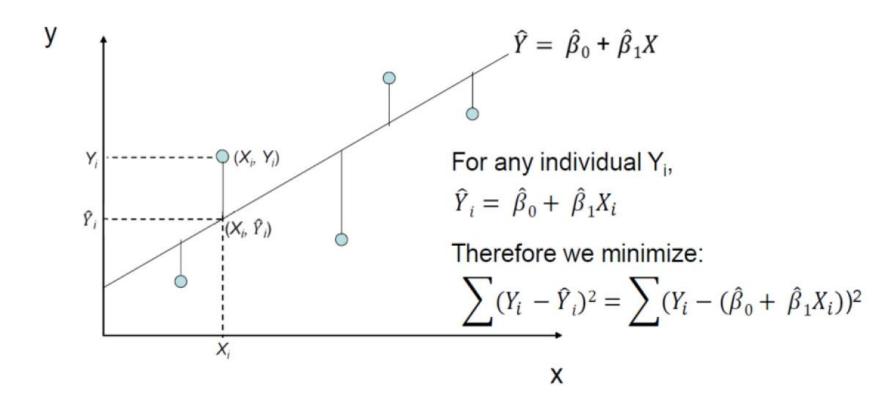
MACHINE LEARNING CODE WITH R





Linear regression

Geometry of Least Square







What is Logistic Regression?

- Many situations where the dependent variable is binary or categorical.
- ✓ Dead vs. Aliv
- ✓ CVD vs No CVD
- ✓ Employed vs. Unemployed
- ✓ Guilty vs. Not guilty
- Requires to consider a special type of model called logistic regression





A new Regression Model

- ➤ Need to modify our regression equation so that:
 - Predictions lie between 0 and 1
 - Effects of covariates can be interpreted on a relative (multiplicative) scale.
- Solution
 - > Use the log odds or logit of p to represent the
 - > Outcome: logit $p = ln(\frac{p}{1-p})$
- Where p is the probability of having the outcome
 - ➤ When p \rightarrow 0, logit p \rightarrow ∞
 - ▶ When $p \rightarrow 1$, logit $p \rightarrow \infty$





A new Regression Model

Logit transformation of the binary
Outcome leads to Logistic Regression

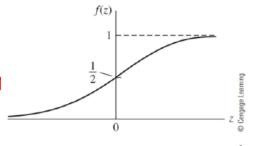


FIGURE 22.1 The logistic function
$$f(z) = \frac{1}{1 + e^{-z}}$$

- The logistic function: $\frac{1}{1+e^{-x}}$
- ightharpoonup Inverse is called the logit: $\ln(\frac{x}{1-x})$
- Logistic regression models:
 - The log-odds (or logit) of a binary outcome as a straightline function of covariates:

$$E[\ln(\frac{p}{1-p})|X] = \beta_0 + \beta_1 * X_1 + ... + \beta_k * X_k$$





Logistic Function

➤ How do we transform the logit back to p?

Logit(p) =
$$\ln\left(\frac{p}{1-p}\right) = \beta_0 + \beta_1 X$$

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

$$\frac{p}{1-p}=e^{(\beta_0+\beta_1X)}$$

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

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

$$p\left(1+e^{(\beta_0+\beta_1X)}\right)=e^{(\beta_0+\beta_1X)}$$

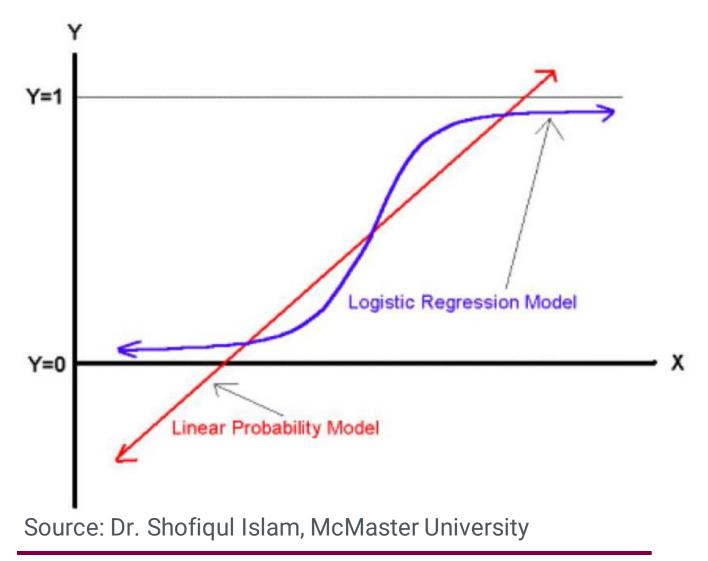
$$p = e^{(\beta_0 + \beta_1 X)} - pe^{(\beta_0 + \beta_1 X)}$$

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





Comparing the Linear vs Logistic fit







Assumptions of Logistic Regression

- ➤ The logistic regression model assumes:
 - > Outcome is a binary or dichotomous variable
 - There is a linear relationship between the logit of the outcome and each predictor variables
 - There is no influential values (extreme values or outliers) in the continuous predictors
 - There are no high correlations (multi-collinearity) among the predictors





Logistic Regression

- Goodness of fit is examined using
 - Measures of predictive ability:
 - ➤ Pseudo R² by McFadden (1974)
 - ➤ Generalized R² by Cox-Snell (1989)
 - ➤ Tjur (2009) coefficient of discrimination
 - ➤ Diagnostic test criteria sensitivity/specificity, area under the ROC curve
- Goodness of fit statistics
 - Deviance and Pearson chi-squared statistics
 - Hosmer-Lemeshow (1980) test
- ➤ Information criteria Akaike (AIC) & Bayesian (BIC)





Machine learning

Machine learning involves showing a large volume of data to a machine/model so that it can learn and make predictions, find patterns, or classify data.





Types of machine learning

Basically, machine learning are three types.







Supervised

Unsupervised

Reinforcement learning.



Source: coursera.org



Supervised learning

Machine learning feeds historical input and output data in machine learning algorithms, with processing in between each input/output pair that allows the algorithm to shift the model to create outputs as closely aligned with the desired result as possible.

Common algorithms used during supervised learning include linear regression, neural networks, decision trees, and support vector machines.



Source: coursera.org



Unsupervised learning

While supervised learning requires users to help the machine learn, unsupervised learning algorithms do not use the same labeled training sets and data. Instead, the machine looks for less obvious patterns in the data.

Unsupervised machine learning is very helpful when you need to identify patterns and use data to make decisions.

Common algorithms used in unsupervised learning include k-means clustering, and Gaussian mixture models.



Source: coursera.org



Reinforcement learning

Reinforcement learning is the closest machine learning type to how humans learn.

The algorithm used learns by interacting with its environment and getting a positive or negative reward.

Common algorithms include temporal difference, and Q-learning.







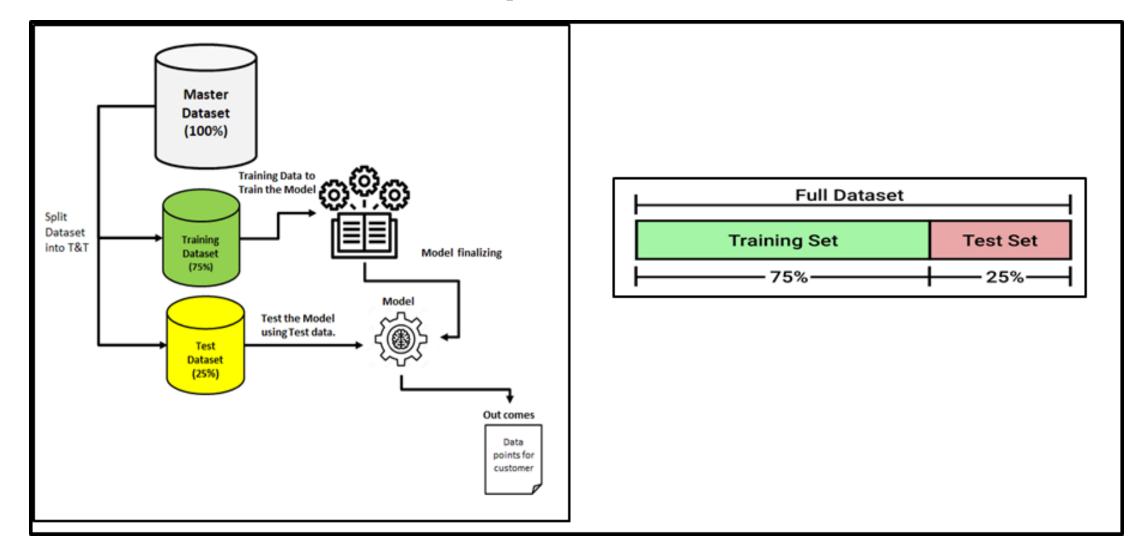
Logistic regression

- A supervised machine learning
- Learns from labeled data
- Make predictions on unseen data





Steps of ML



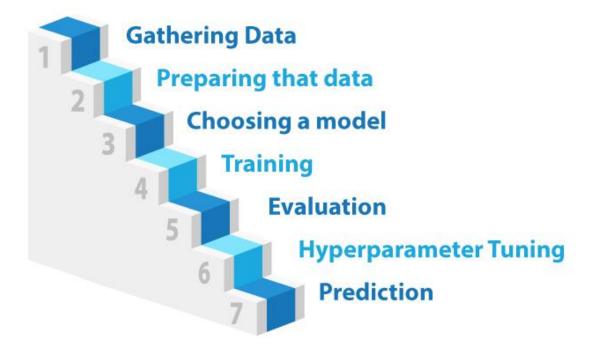




Library

Steps of ML including tuning







www.mygreatlearning.com



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Let move to the coding part

https://colab.research.google.com/drive/1iOv C52mkSdQ-EkNQZFSi-O9y05YSRSMs#scrollTo=yi7vfdCrHQaU



