

Data Classification using K-NN: Personal Loan Acceptance Case-Study

1) Analytic Problem

Universal Bank is a relatively young bank growing rapidly in terms of overall customer acquisition. The majority of these customers are liability customers (depositors) with varying sizes of relationship with the bank. The customer base of asset customers (borrowers) is quite small, and the bank is interested in expanding this base rapidly to bring in more loan business. In particular, it wants to explore ways of converting its liability customers to personal loan customers (while retaining them as depositors).

A campaign that the bank ran last year for liability customers showed a healthy conversion rate of over 9% success. This has encouraged the retail marketing department to devise smarter campaigns with better target marketing. The goal of our analysis is to model the previous campaign's customer behavior to analyze what combination of factors make a customer more likely to accept a personal loan. This will serve as the basis for the design of a new campaign.

2) Data Exploration

The file UniversalBank.xls contains data on 5000 customers. The data include customer demographic information (age, income, etc.), the customer's relationship with the bank (mortgage, securities account, etc.), and the customer response to the last personal loan campaign (*Personal Loan*). Among these 5000 customers, only 480(= 9.6%) accepted the personal loan that was offered to them in the earlier campaign.

3) Model Construction & Prediction

Consider the following customer:

Age=40, Experience=10, Income=84, Family=2, CCAvg=2, Education_2=1,
Education_3=0, Mortgage=0, Securities Account=0, CD Account=0, Online=1 and
Credit card=1.

3.1) Perform a k -NN classification with all predictors except ID and ZIP code with using $k = 1$.

3.2) How would this customer be classified?

4) Improving Performance of Classifier

- 4.1) Partition the data into training (60%) and validation (40%) sets. Show the classification matrix for the validation data that results by varying k .
- 4.2) Using the best k , how would this customer be classified?
- 4.3) Extra-points => Redo all tasks using Python or R