

Machine Learning

Dr Changjiang He, Dr Kuo-Ming Chao
Computer Science| School of Art
University of Roehampton

Lesson 4.1

Evaluation of Prediction

- Linear Regression
- Lasso
- Logistic Regression
- Random Forest
- Neural Network (next week)
- ...

Root Mean Square Error

RMSE is a popular formula to measure the error rate of a regression model.

However, we can only compare between models whose errors we can measure in the same units

$$RMSE = \sqrt{\frac{\sum_{i=1}^n (p_i - a_i)^2}{n}}$$

Relative Square Error

Unlike RMSE, the relative squared error (RSE) can be compared between models whose errors we can measure in different units

$$RSE = \frac{\sum_{i=1}^n (p_i - a_i)^2}{\sum_{i=1}^n (\bar{a} - a_i)^2}$$

Relative Absolute Error

Like RSE, the relative absolute error (RAE) can be compared between models whose errors are measured in the different units.

$$RAE = \frac{\sum_{i=1}^n |p_i - a_i|}{\sum_{i=1}^n |\bar{a} - a_i|}$$

Mean Absolute Error

Mean Absolute Error is the average of the difference between the Original Values and the Predicted Values. It gives us the measure of how far the predictions were from the actual output. However, they don't give us any idea of the direction of the error i.e. whether we are under predicting the data or over predicting the data. Mathematically, it is represented as :

$$MeanAbsoluteError = \frac{1}{N} \sum_{j=1}^N |y_j - \hat{y}_j|$$

Coefficient of Determination

The coefficient of determination (R^2) summarizes the explanatory power of the regression model and is computed from the sums-of-squares terms.

R^2 describes the proportion of variance of the dependent variable explained by the regression model. If the regression model is “perfect”, SSE is zero, and R^2 is 1. If the regression model is a total failure, SSE is equal to SST, no variance is explained by regression, and R^2 is zero.

Coefficient of Determination → $R^2 = \frac{SSR}{SST} = 1 - \frac{SSE}{SST}$

Sum of Squares Total → $SST = \sum (y - \bar{y})^2$

Sum of Squares Regression → $SSR = \sum (y' - \bar{y}')^2$

Sum of Squares Error → $SSE = \sum (y - y')^2$

- These density plots show the predicted vs. actual values for the regression models to predict the the response variable.
- Note: the yellow/lightest areas represent the highest density of points

