

# 02 Model Diagnostics

## Bias-Variance Tradeoff

Predictability measure (error on test set) = variance + bias

## Evaluate Goodness of Model

### For Linear Regression/ GLM

We can use hypothesis test and model selection criteria to evaluate the goodness of a LR/ GLM model.

#### Hypothesis Test

- Wald test
- Score test
- Likelihood ratio test

#### Model Selection Criteria

- AIC
- BIC
- ...

### For all Type of Supervised Model

#### Training Set / Validation Set / Test Set

- **Training set:** used to train a model
- **Validation set:** used to evaluate the performance of trained models, their performance will be used to model selection
- **Test set:** pass the test set to your final model, the resulting performance (eg. accuracy) is the final performance.

#### Cross-Validation

Randomly split the whole dataset to p equal-sized subset. For each subset, use it as validation set (for model selection) and use other p-1 subset as training set to get a model. We will get p models in total, and p validation set error in total. The final performance is the average of the p validation set error.

#### Gains Curve

Tools to evaluate models performance in context of regression with **continuous response**.

**Gain curve at about is better.**

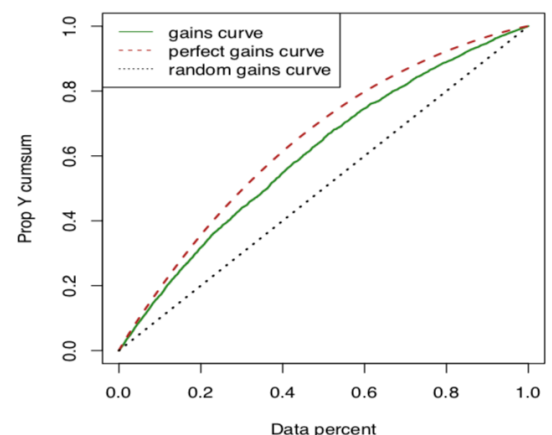
Y轴:  $\frac{\sum_{i=0}^j Y_i}{\sum_{k=1}^n Y_k}$  for  $j = 1 \dots n$  in case of  $(f(X_i), Y_i)$  pairs are sorted by the value of  $f(X_i)$  from high to low

X轴:  $\frac{j}{n}$  for  $j = 1 \dots n$

- 灵感来自于low predictions should corresponding with low response observations, high predictions should corresponding with high observations.
- 比如现在有两个模型  $f_1(X)$ ,  $f_2(X)$  和一个数据集 of observations  $X_i$ s.
- Sort  $(f_1(X_i), Y_i)$  pairs by the value of  $f_1(X_i)$  from high to low.
- Look how good the  $Y_i$ 's are in a sorted sequence? 按照predictions sorting之后, true response是不是也被sorted了? 如果没有, 说明predictions错得厉害.
- Sorting for  $f_2(X)$ , 看看哪个sorting 过程能把true response Y排序得好.
- 把Y排得好的那个model为佳.

**Note:**

- 当observation很多时, 很难看出Y sort得好不好, 所以使用  $\frac{\text{cumulative sum of } Y}{\text{total sum of } Y}$  to represent the proportion of cumulative Y. It is better if the term increase earlier.
- So for Gain curve, **above is better**.
- Gain curve 只关注顺序. 忽略了具体值. 不能保证goodness of fit.



All has perfect gain chart, but...

| $Y_i$ | $f_1(X_i)$ | $f_2(X_i)$ | $f_3(X_i)$ | $f_4(X_i)$ |
|-------|------------|------------|------------|------------|
| 1     | 1          | 107        | 2.98       | -1,000     |
| 2     | 2          | 102        | 2.99       | 0          |
| 3     | 3          | 103        | 3.00       | 0.1        |
| 4     | 4          | 104        | 3.01       | 50         |
| 5     | 5          | 105        | 3.02       | 50,001     |

## ROC Curve (Receiver Operating Characteristic Curve) & AUC

ROC curve is a tools (which is similar as gain curve) to evaluate models performance in context of **binary response** {0,1}. AUC is the area under the ROC curve

Y轴: Sensitivity rate for variant cutoff

X轴: False Positive rate for variant cutoff

- Sensitivity Rate: proportion of real 1 predicted as 1
- False negative Rate: proportion of real 1 predicted as 0
- Specificity Rate: proportion of real 0 predicted as 0
- False Positive Rate: proportion of real 0 predicted as 1

### Note:

左上角的ROC is best, 对应AUC的最大值1.0

