

Machine Learning @MIT Lecture Note 11 notes 1

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Question 1.

Lecture note 11 Model selection criteria: Minimum description length and Feature selection.

Some note from Lecture note 11 @credit to MIT

Model selection criteria - compressed data - classification context - cost of error in training - cost of selected classifier

Simple two-part MDL $P(y_t) = (y_t, 1)(1)(y_t, 1 - \text{minimize encoding cost } \min \log P(y_t) = \log P(y_t^*) (4) || t=1 \dots n = l(y_1, \dots, y_n; \hat{\theta})$ - Find distribution in model \rightarrow minimize encoding length of data - Uniform distribution over possible discrete values $DL_{\text{of data given model}} DL_{\text{of model}} \quad DL = l(y_1, \dots, y_n; \hat{\theta}) + \log(n+1)$

- description length of sequence

Universal coding, normalized maximum likelihood

- Find distribution in closest to just encoding the data with the best fitting distribution in model $\log P_{\text{NML}}(y_1, \dots, y_n) \min_l(y_1, \dots, y_n; \theta) = \max_l(y_1, \dots, y_n; \theta)$ - Distribution minimize maximum deviation \rightarrow normalized maximum likelihood distribution - minimax optimal coding length $\log P_{\text{NML}}(y_1, \dots, y_n) = \max_l(y_1, \dots, y_n; \theta) + \log \exp \max_l(y_1, \dots, y_n; \theta) (1)$

FEATURE SUBSET SELECTION - Feature vector \rightarrow useful for classification - deal with noisy and irrelevant feature data - weight how useful data - feature weighting problem - Kernel optimization - Naive bayes, assume none of feature and label independent - log likelihood - ML param estimate - Shannon entropy - Conditional entropy - log-likelihood if we assume feature to depend on label \rightarrow mutual infor

Mutual information