

SNS COLLEGE OF TECHNOLOGY

Coimbatore-35 An Autonomous Institution

Accredited by NBA – AICTE and Accredited by NAAC – UGC with 'A++' Grade Approved by AICTE, New Delhi & Affiliated to Anna University, Chennai

DEPARTMENT OF ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING 23AMB201 - MACHINE LEARNING

II YEAR IV SEM

UNIT II – SUPERVISED LEARNING ALGORITHM

TOPIC 1 – Bias and variance Overfitting Underfitting

Redesigning Common Mind & Business Towards Excellence











Build an Entrepreneurial Mindset Through Our Design Thinking FrameWork



Regression problem: Underfitting, overfitting, best fitting

- **Underfitting:** A model or a ML algorithm is said to have underfitting when it cannot capture the underlying trend of the data.
- _2_ Overfitting: A model or a ML algorithm is said to have overfitting when it capture the
- underlying trend of the data very accurately. 3
 - **Best fit:** A model or a ML algorithm is said to have best fit when it capture the underlying trend of the data moderately.







Bias: It shows the degree of randomness of the training data.

- Based on the training data a suitable model may be created for the regression (a) or classification problem.
- **Regression:** Linear, logistic, polynomial (b)

(c) Classification: Decision tree, random forest, naive baye's, KNN, SVM Variance: It shows the degree of randomness of the testing data. Testing data validates the accuracy of a model, that has been made with the **(i)**

- help of training data set.
- (ii) Testing data is nothing but the unlabeled or unknown data.





Underfitting, overfitting, best fit and bias, variance



Note:

- \Rightarrow The objective of ML algorithm not only fit for the training data but also fit for the testing data.
- In other words, low bias and low variance is the appropriate solution. \Rightarrow

Underfitting	Overfitting	
High bias	Very low bias	L
High variance	High variance	L



Best fit

ow bias

ow variance

A classifier (Decision tree, random forest, naive baye's, KNN, SVM) works on two types of data

- Training data Testing
- data

Example:

Classifier comes under the category of underfitting, overfitting, and best fit based on its training and testing accuracy.

Underfitting	Overfitting
Train error=25%	Train error=1%
Test error= 27%	Test error= 23%



Best fit Train error=8% Testerror=9%



Mathematical intuition



bias
$$\hat{f}(x) = E[\hat{f}(x)] - f(x)$$
 (
variance $\hat{f}(x) = E^{h} \hat{f}(x) - E[\hat{f}(x)]^{2^{i}}$ (2)

- $\Rightarrow \hat{f}(x) \rightarrow \text{output observed through the training model}$
- ⇒ For linear model $\hat{f}(x) = w_1 x + w_0$ ⇒ For complex model $\hat{f}(x) = \sum_{i=1}^{D} w_i x^i + w_0$
- \Rightarrow We don't have idea regarding the true f(x).
- ⇒ Simple model: Low bias & high variance
- ⇒ Complex model: High bias & low variance

 $E[(y - f(x))^2] = bias^2 + Variance + \sigma^2 (Inreducible error)$



1) 2)

(3)



and variance trade-off relation











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