Machine Learning In Python

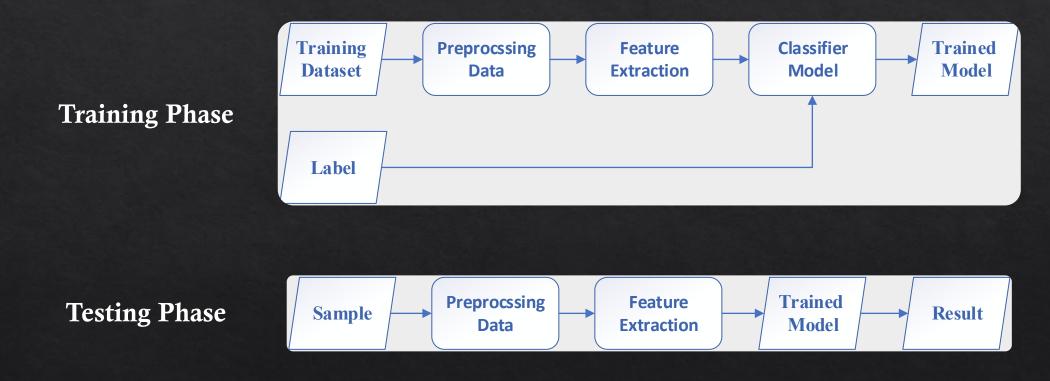
Subject: Classification Using Naïve Bayes, Overfitting vs. Underfitting

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Classification In Supervised Learning Framework



Classifier Models:

Naïve Bayes

- Naïve Bayes is a supervised classification algorithm which uses the Bayes' theorem for classification performance.
- Naïve Bayes classifiers are a family of simple "probabilistic classifiers" based on applying Bayes' theorem with strong (naïve) independence assumptions between the features.

Bayes' theorem

- $P(C_k|X)$: Posterior Probability
- $P(X|C_k)$: Likelihood
- P(X): Prior Probability

$$P(C_k|X) = \frac{P(C_k)P(X|C_k)}{P(X)}$$
$$X = [x_1, x_2, ..., x_d]$$

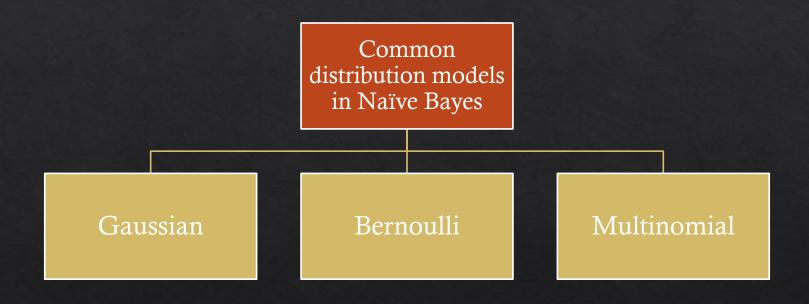
How to train and test a Naïve Bayes classifier?

- Suppose we have a binary classification problem (k=1, 2).
- Training:
 - Two probability distributions is fitted to the training samples of each class $(P(C_1|X), P(C_2|X))$.
- Testing:
 - The testing sample is given to each probability distribution in order to generate probabilities belongs to each class $(P(C_1|X_T), P(C_2|X_T))$.
 - $P(C_1|X_T) > P(C_2|X_T) \rightarrow X_T$ is classified to the C_1 by Naïve Bayes model
 - $P(C_2|X_T) > P(C_1|X_T) \rightarrow X_T$ is classified to the C_2 by Naïve Bayes model

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Probability Distribution Models in Naïve Bayes



Overfitting vs. Underfitting

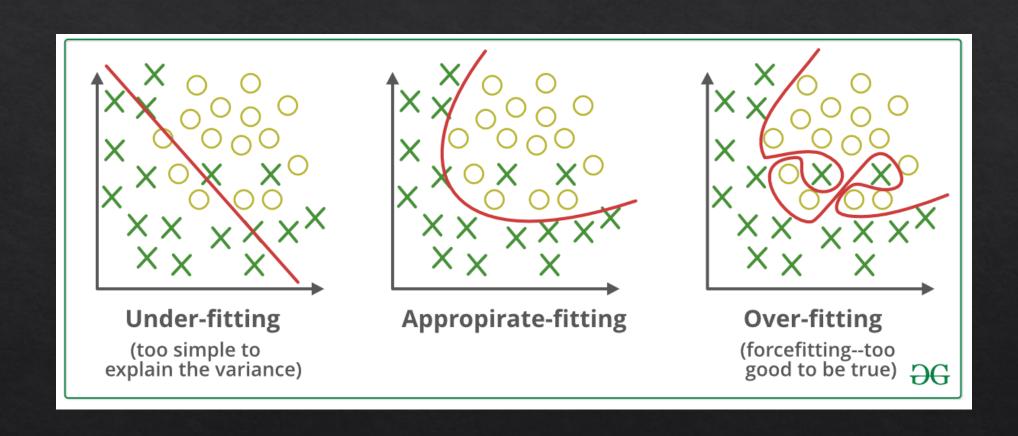
Overfitting:

- Overfitting refers to a model that models the training data too well.
- Overfitting happens when a model learns the detail and noise in the training data to the extent that it negatively impacts the performance of the model on new data.
- This means that the noise or random fluctuations in the training data is picked up and learned as concepts by the model.
- The problem is that these concepts do not apply to new data and negatively impact the models ability to generalize.

Underfitting:

- Underfitting refers to a model that can neither model the training data nor generalize to new data.
- An underfit machine learning model is not a suitable model and will be obvious as it will have poor performance on the training data.

Overfitting vs. Underfitting



Overfitting vs. Underfitting

The main reasons of causing Overfitting:

- The high number of features of the dataset.
- Simplicity of data.
- The high number of training samples

