

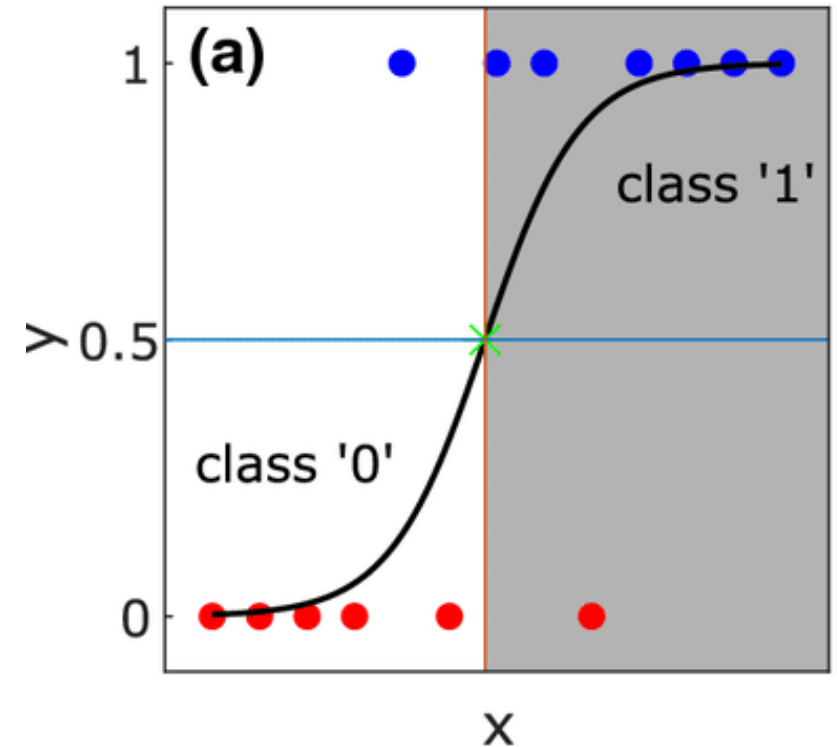
Computer Vision

sec4

Logistic Regression

* Logistic regression is a statistical model used for classification tasks, designed to predict the probability that a given input belongs to a particular class. It can handle both **binary classification** (where there are two possible outcomes) and **multi-class classification** (where there are more than two classes).

* The model estimates probabilities using a logistic (sigmoid) function for binary cases, or a softmax function for multi-class cases



important Parameters:

- C: Inverse of regularization strength (smaller values specify stronger regularization).
- solver: Algorithm to use in the optimization problem
(e.g., 'lbfgs', 'liblinear', 'newton-cg', 'saga', 'sag').
- max_iter: Maximum number of iterations for the solver.
- Penalty: type of regularization (l1, l2)

Decision Tree

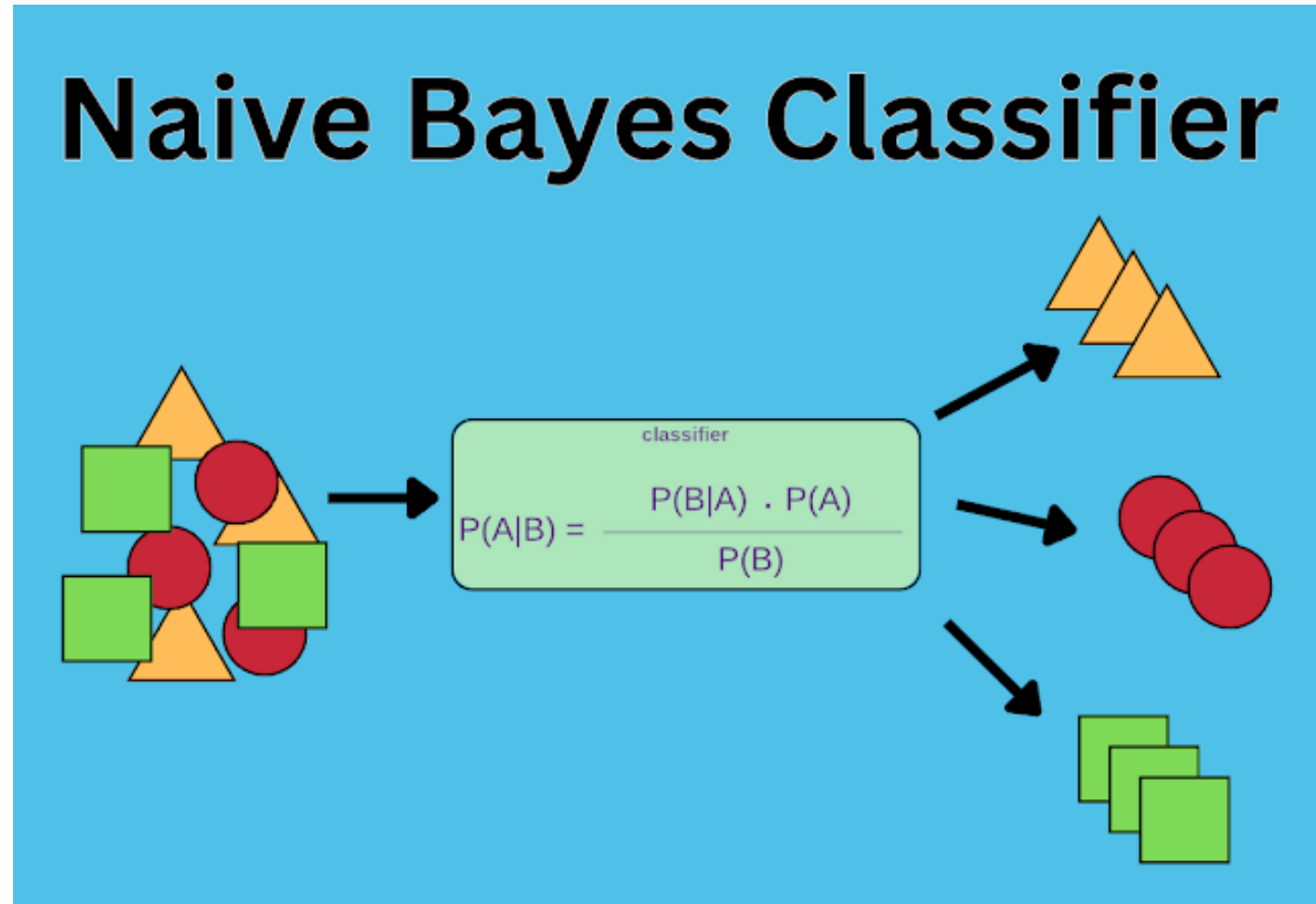
Decision Tree is a supervised learning method used for classification. They split the data into subsets based on the value of input features, creating a tree-like model of decisions.

important Parameters:

- `criterion`: Function to measure the quality of a split ('gini' , 'entropy').
- `max_depth`: Maximum depth of the tree (limits the number of splits).

Naïve Bayes

Naive Bayes is a probabilistic algorithms based on Bayes' theorem, which is used for classification tasks.



Important parameters

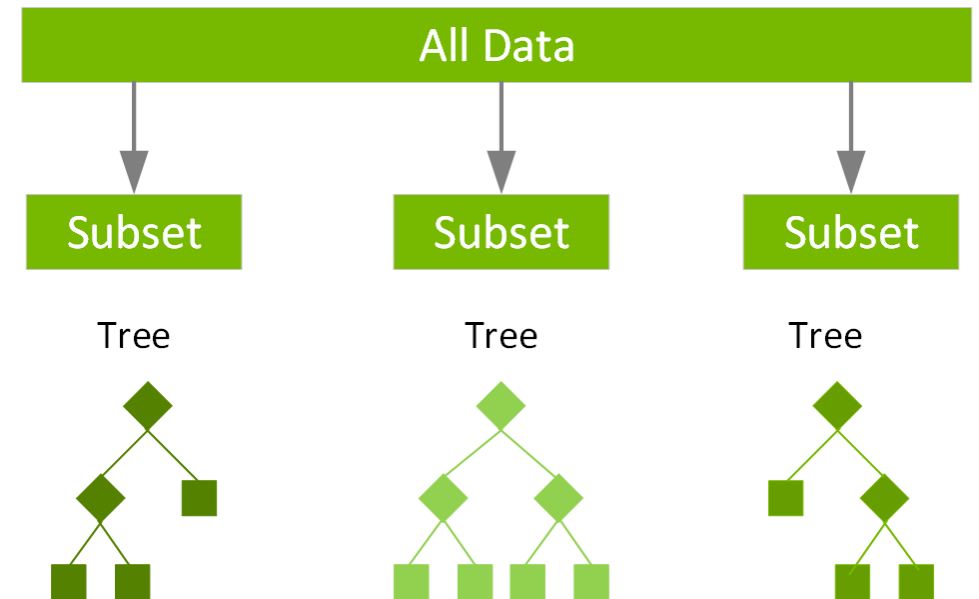
- `var_smoothing`: A small constant added to feature variances to ensure numerical stability during calculations. Helps avoid issues when variances are too small or zero.

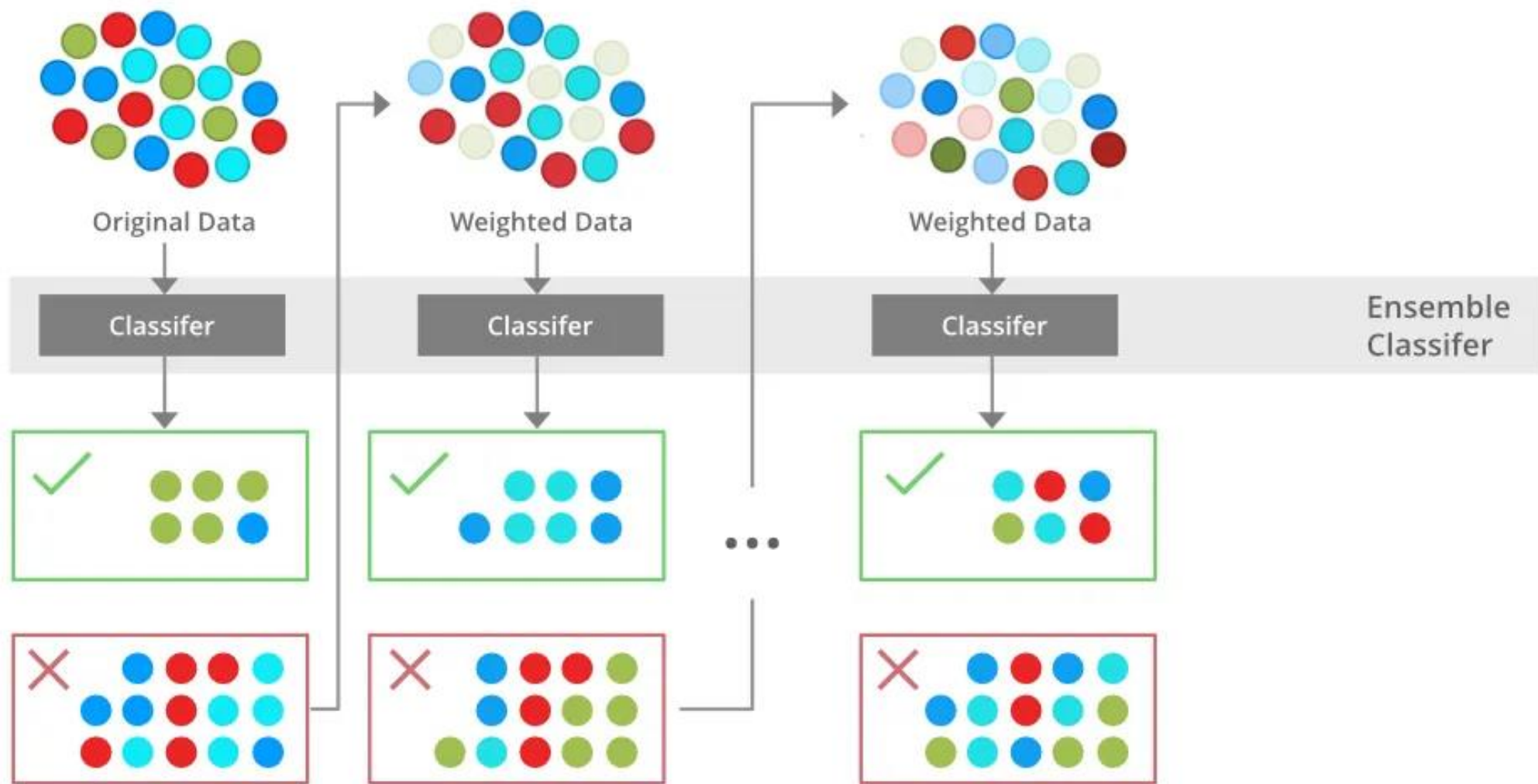
XGBoost

XGBoost (Extreme Gradient Boosting) is a powerful and efficient open-source machine learning library designed for gradient boosting

XGBoost implements the gradient boosting algorithm, which combines the predictions of multiple weak learners (typically decision trees) to create a strong predictive model.

It builds trees sequentially, with each new tree aiming to correct the errors made by the previously built trees.



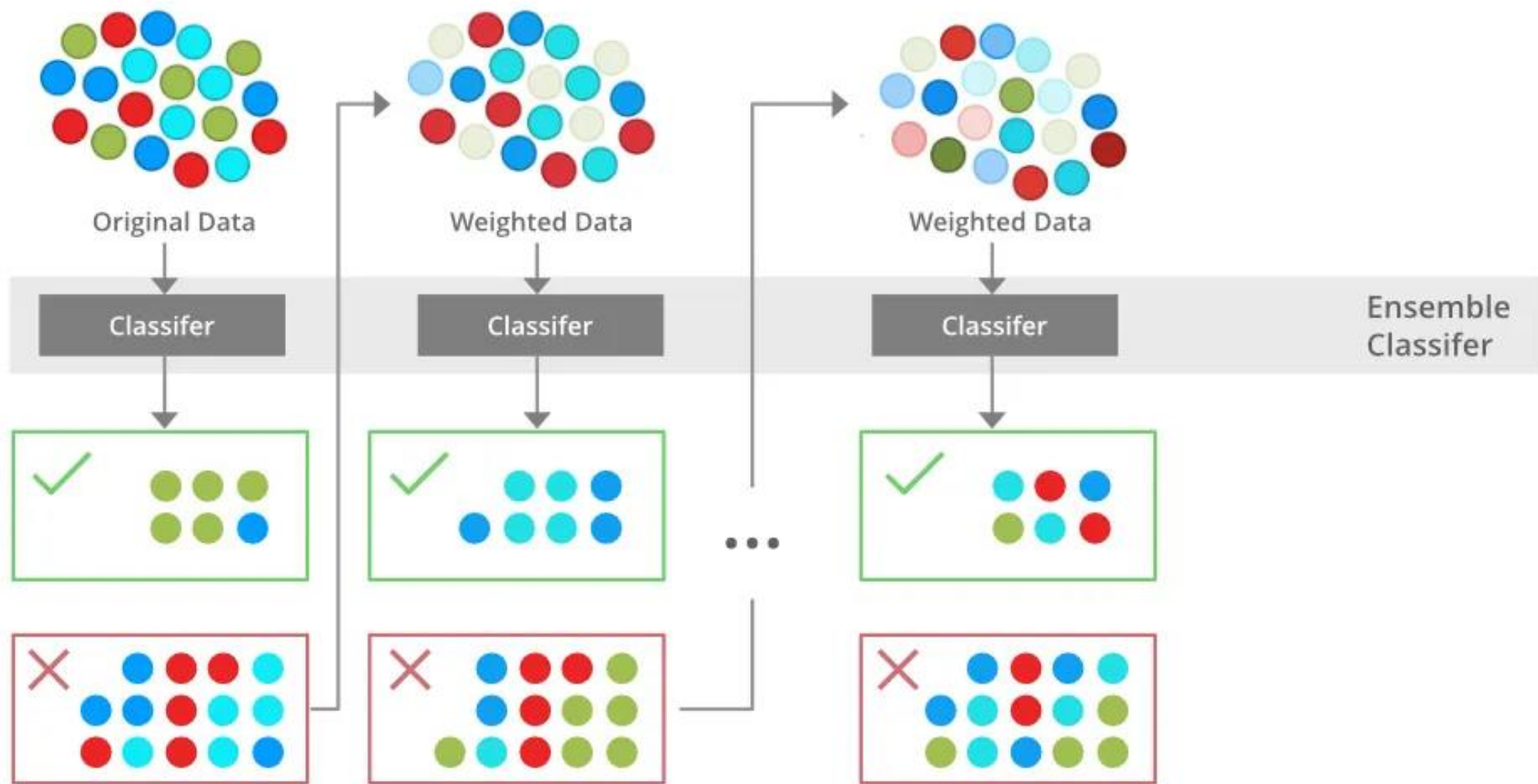


Important parameters

- **n_estimators**: The number of boosting rounds (trees to build)
- **learning_rate**: used to prevent overfitting. A lower value makes the model more robust
- **max_depth**: maximum depth of the individual trees
- **eval_metric**: The default evaluation metric varies based on the task (e.g., logloss for binary classification, mlogloss for multi-class).

CatBoost

- Boosting algorithm
- It builds trees sequentially, with each tree correcting the errors of the previous ones, similar to other boosting algorithms like XGBoost
- It uses a technique called **order-based encoding** that allows it to efficiently incorporate categorical data into the model.
- **Output:** Final prediction is made through a aggregation



Important parameters

- **Iterations:**The total number of trees
- **learning_rate:**Lower values can lead to better generalization
- **Depth:**maximum tree depth

Adaboost

- Boosting algorithm
- Combines multiple weak learners (often decision trees) in a sequential manner. Each weak learner is trained to correct the errors made by the previous ones, focusing on misclassified examples by adjusting their weights.
- **Output:** Final prediction is made through a majority vote

Important parameters

- **Estimator:** base estimator
- **n_estimators:** number of trees
- **learning_rate:** A lower value makes the learning process more robust
- **Algorithm:** SAMME, SAMME.R

LightGBM

- is an open-source, distributed gradient boosting framework developed by Microsoft. It is designed to be efficient and scalable for machine learning tasks, particularly for large datasets.

Important parameters

- **boosting_type**:gbdt, dart, goss
- **num_leaves**:maximum number of leaves in one tree
- **max_depth**:maximum depth of the tree
- **learning_rate**: A lower value makes the learning process more robust
- **n_estimators**:number of trees
- **Objective**:binary, multiclass