

Bagging (Bootstrap Aggregating) – Introduction

1. Core Idea

- Bagging is short for **Bootstrap Aggregating**.
- It's an ensemble method that combines multiple models to improve accuracy and stability.
- Works by:
 - 1. Taking **random subsets** of the training data (with replacement → bootstrap).
 - 2. Training a separate model on each subset.
 - 3. Combining predictions:
 - Classification → majority vote.
 - **Regression** → average.

2. Why Bagging?

- Single models (like decision trees) can be very unstable → small changes in data cause big differences in predictions.
- Bagging reduces variance by averaging results of many models → making predictions more stable and accurate.

3. Process (Step by Step)

Imagine you have a dataset with 100 rows.

1. Bootstrap Sampling:

- 1. Randomly pick 100 rows with replacement → some rows repeat, some are left out.
- 2. Do this multiple times to create several training sets.

2. Model Training:

- 1. Train a model (say a decision tree) on each sample.
- 2. Each model learns differently because each sample is slightly different.

3. Aggregation:

- Combine outputs of all models:
 - Classification → majority voting.
 - Regression → averaging.

4. Example

Suppose we want to classify whether an email is spam:

- Model 1 says: Spam
- Model 2 says: Not Spam
- Model 3 says: Spam
- Final (Majority Vote) → Spam

5. Advantages of Bagging

- 1. Reduces **variance** → avoids overfitting.
- 2. Works well with high-variance models (e.g., Decision Trees).
- 3. Easy to implement.

6. Disadvantages

- X Not effective if the base model already has **low variance** (like linear regression).
- X More computationally expensive (training many models).

7. Intuition (Easy Analogy)

Think of guessing a jar's number of candies:

- If you ask one person → they might be far off.
- If you ask 100 people and take the average → the result is usually much closer to the truth.

That's bagging → "wisdom of the crowd" in action.

❖ Final Takeaway:

- Bagging = **Bootstrap sampling + Aggregation**.
- Reduces variance & increases stability.
- Often used with decision trees → leads to Random Forest (which is bagging + feature randomness).