



# Hyperparameter Tuning



## What is Hyperparameter Tuning?

In Machine Learning, there are **two types of parameters**:

### 1. Model Parameters (Learned Automatically)

These are learned from data during training.

Example:

- Weights in Linear Regression
- Coefficients in Logistic Regression
- Node splits in Decision Trees

**You do NOT set these manually**



# Hyperparameter Tuning



## 2. Hyperparameters (Set by Us Before Training)

These control **how the model learns**.

Examples:

- `k` in KNN
- `max_depth` in Decision Tree
- `n_estimators` in Random Forest
- `learning_rate` in Gradient Boosting

These **must be chosen carefully**

Wrong values = Bad model performance

Hyperparameter Tuning = Finding the BEST combination of hyperparameters for maximum model performance.



# Hyperparameter Tuning



## Why Do We Need Hyperparameter Tuning?

Because:

- Too simple model → **Underfitting**
- Too complex model → **Overfitting**

Correct hyperparameters → **Best Generalization**

Goal: **Maximize accuracy on unseen data**



# Hyperparameter Tuning: Types



## 1) GridSearchCV:

GridSearchCV is an **automated hyperparameter tuning method** that:

- Tries **all possible combinations** of hyperparameters
- Uses **Cross-Validation (CV)**
- Finds the **best performing model**



# Hyperparameter Tuning: 2 Main Types



## 2) RandomizedSearchCV

Instead of testing all combinations, it **randomly selects some combinations**.

### Pros:

- Much faster than GridSearch
- Works well for large parameter spaces
- Finds near-optimal models quickly

### Cons:

Does not guarantee optimal result, since it uses Random Search

**Best used when you have** large datasets, deep models or limited computation time



Fhdsklf  
Fjdsklf  
Fjsklfd

# Heading Goes Here

