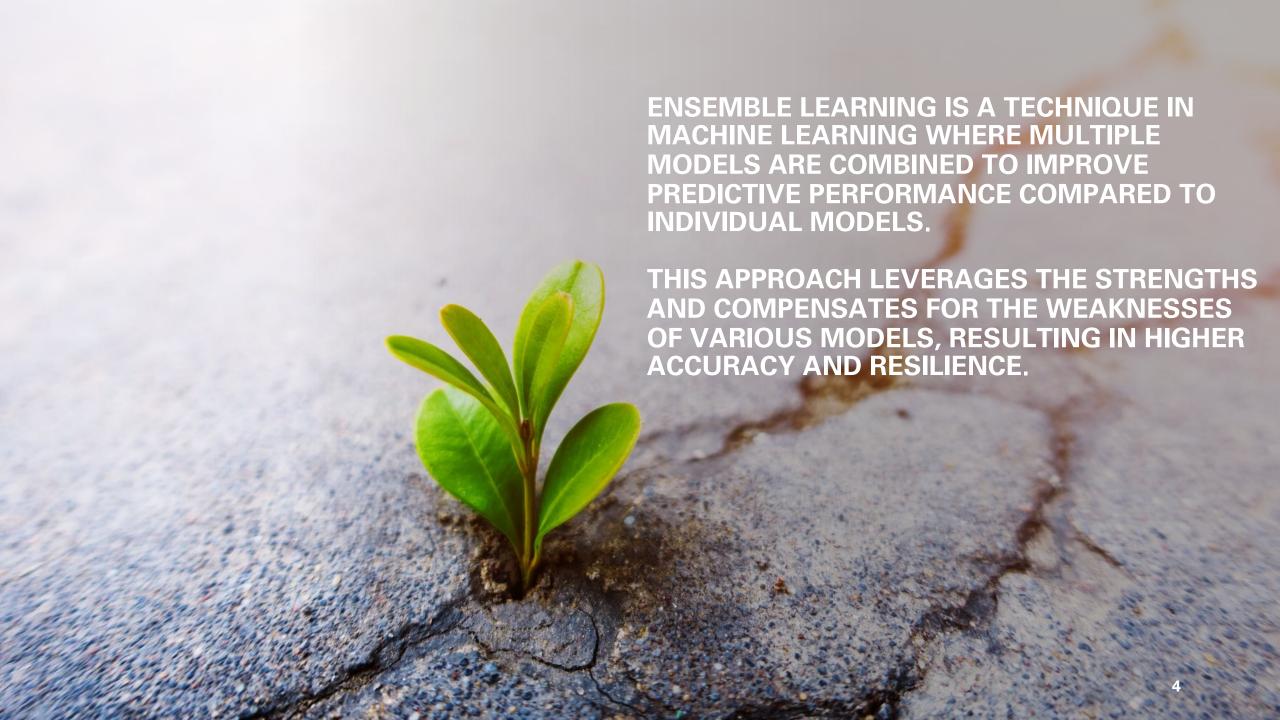


ENSEMBLE LEARNING

What is ensemble learning? Type of ensemble learning Comparison Further reading





Type of ensemble learning







Boosting



Stacking



Comparison



	Bagging	Boosting	Stacking
Purpose	Reduce variance	Reduce bias	Combine multiple models to improve predictive performance
Combination method	Averaging or voting	Weighted voting	Meta-learner (stacker) combines predictions of base models
Performance	Improves stability and accuracy by averaging out noise	Improves accuracy by focusing on difficult-to-predict instances	Often outperforms individual models by leveraging their strengths
Model training	Trains models independently	Trains models sequentially, each focusing on errors of the previous	Trains models independently first, then trains a meta-learner on their predictions
Handling noise	More robust to noisy data	Sensitive to noisy data and outliers	Can mitigate overfitting by using a diverse set of base models

Comparison



	Bagging	Boosting	Stacking
Complexity	Easier to implementsimpler computation	More complex implementationhigher computational cost	 Most complex to implement requires careful selection training of base models and meta-learner
Overfitting	Less prone to overfitting	Can overfit, especially with noisy data	Prone to overfitting if base models or meta-learner are not properly regularized
Weak learners	Can use any models as base learners, including strong and weak learners	Uses weak learners and iteratively improves them	Can use any models as base learners, including strong and weak learners
Hyperparameters tuning	Requires tuning of fewer hyperparameters	Requires careful tuning of hyperparameters	Requires tuning of both base models and the meta-learner

FURTHER READING

Machine Learning & Pattern Recognition Series – Ensemble Learning Foundations and Algorithms + o

THANK YOU