Topic 20: Random Forest

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Key points: .

Disclaimer: *The note is built on Prof. Jinchi Lv's lectures of the course at USC, DSO 607, High-Dimensional Statistics and Big Data Problems.*

20.1   Motivation

Denote by $m(\mathbf{X})$ the measurable nonparametric regression function with p-dimensional random vector $\mathbf{X}$ taking values in $[0, 1]^p$. The Random Forest algorithm aims to learn the regression function in a nonparametric way based on the observations $\mathbf{x}_i \in [0, 1]^p$, $y_i \in \mathbb{R}$, $i = 1, \cdots, n$, from the model

$$y_i = m(\mathbf{x}_i) + \epsilon_i$$

where $\mathbf{X}, \mathbf{x}_i, \epsilon_i, i = 1, \cdots, n$ are independent, and $\{\mathbf{x}_i\}$ and $\{\epsilon_i\}$ are two sequences of identically distributed random variables. $\mathbf{x}_i$ is distributed identically as $\mathbf{X}$.

Why Random Forest (RF)?   **RF** has gained significant popularity due to its

- **High accuracy**: **RF** consistently rank among the top performer, often surpassing more complex models
- **Robustness**: **RF** are less subject to overfitting due to the ensemble nature leveraging multiple decision trees
- **Interpretability**: **RF** provide rankings of feature importance

Chi et al. (2022)

References

Chien-Ming Chi, Patrick Vossler, Yingying Fan, and Jinchi Lv. Asymptotic properties of high-dimensional random forests. *The Annals of Statistics*, 50(6):3415–3438, 2022.