

# Interpretable ML

PDP, ICE, ALE

# Partial Dependence Plots

Plotting feature effects in “black box” learning methods (Friedman 2001)

$$\tilde{f}(x) = \frac{1}{n} \sum_{i=1}^n f(x, x_{iC})$$

General idea

- Compute  $\tilde{f}(x)$  over the range of  $x$  while averaging the effects of the remaining predictors  $x_C$
- Generate artificial datasets by fixing  $x$ -values for all cases
  - Regression: Averaging over  $f(x, x_{iC})$  for each value of  $x$
  - Classification: Averaging over  $p$  or  $\text{logit}(p)$  for each value of  $x$

# Partial Dependence Plots

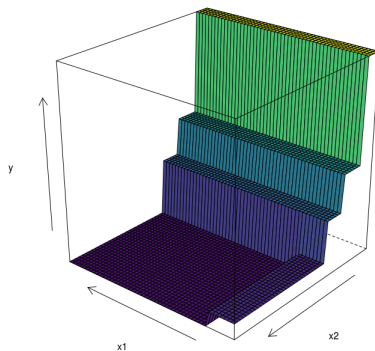
## Constructing PDPs

- ① Choose a range of values  $\{x_{11}, x_{12}, \dots, x_{1k}\}$  of  $x_1$
- ② For each  $i \in \{1, 2, \dots, k\}$ 
  - ① Generate an artificial dataset by fixing  $x_1$  to  $x_{1i}$  for all cases
  - ② Compute predictions for all cases using the prediction model (e.g. RF)
  - ③ Average the predictions over all cases
- ③ Plot the obtained average predictions against  $x_{1i}$  for  $i = 1, 2, \dots, k$

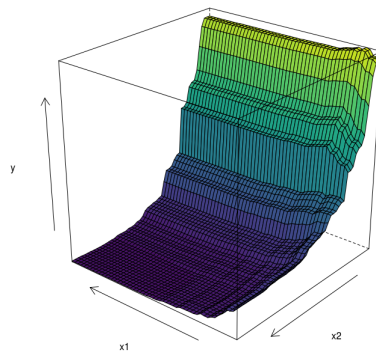
# Partial Dependence Plots

Figure: Partial dependence plots

(a) CART



(b) Random Forest



# Individual Conditional Expectation

ICE plots (Goldstein et al. 2014)

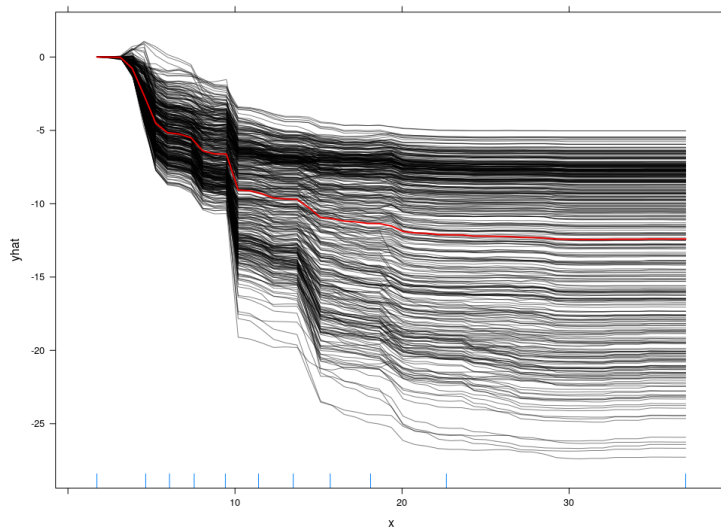
- Individual PDPs for all cases w/o final averaging
- One line represents the predictions for one case over the range of  $x$
- Can uncover heterogeneous effects that are driven by interactions

Centered ICE plots

- Adjusts for different individual baselines
- Shows differences in prediction relative to anchor (e.g.,  $x_{min}$ )

# Individual Conditional Expectation

Figure: Centered ICE plot



# Accumulated Local Effects

## ALE plots (Apley 2016)

- With correlated features, PDPs can (artificially) construct very unlikely combinations
- ALE solution:
  - ① Use only cases with (similar)  $x$ -values within a given interval
  - ② Calculate differences in predictions between upper and lower limit of this interval

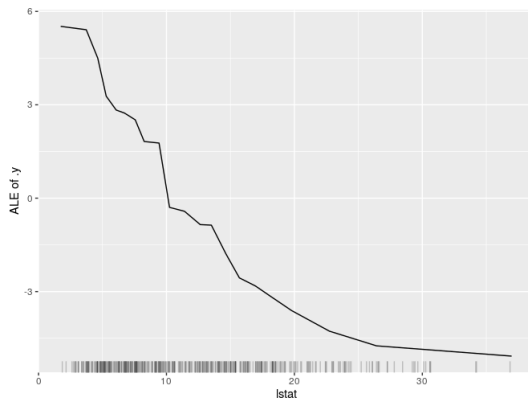
$$\hat{f}_{j,ALE}(x) = \sum_{k=1}^{k_j(x)} \frac{1}{n_j(k)} \sum_{i: x_j^{(i)} \in N_j(k)} \left[ f(z_{k,j}, x_{\setminus j}^{(i)}) - f(z_{k-1,j}, x_{\setminus j}^{(i)}) \right]$$

→ Differences in predictions in interval  $z_{k,j}, z_{k-1,j}$  for cases in neighborhood  $N_j(k)$  accumulated up to interval  $k_j$

# Accumulated Local Effects

Figure: Comparison of feature effect plots

(a) ALE



(b) PDP

