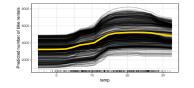
Interpretable Machine Learning

PDP - Comments and Extensions



Learning goals

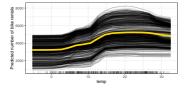
- Extrapolation and Interactions in PDPs
- Centered ICE and PDP



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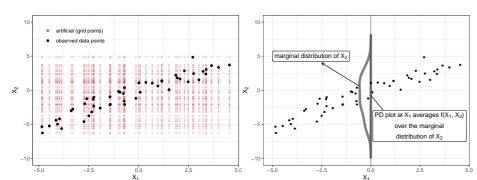
Feature Effects
PDP - Comments and Extensions

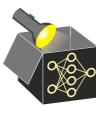


Learning goals

- Extrapolation and Interactions in PDPs
- Centered ICE and PDP

COMMENTS ON EXTRAPOLATION

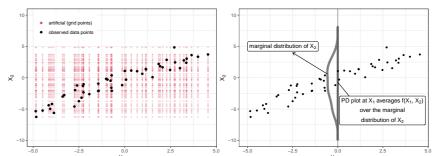




Extrapolation occurs in regions with few observations or if features are correlated

- **Example:** Features x_1 and x_2 are strongly correlated
- Black points: Observed points of the original data
- Red: Grid points to calculate ICE/PD (many unrealistic x_1 , x_2 combinations)
 - \Rightarrow **PD at** $x_1 = 0$: Averages predictions over *full* marginal distribution of x_2
 - ⇒ **Issue:** Model may behave strangely outside training distribution
 - ⇒ Especially problematic for overfitted or interaction-heavy models

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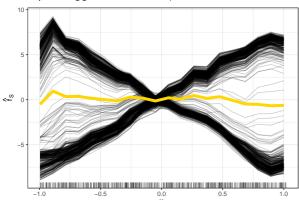
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COMMENTS ON INTERACTIONS

PD plots average ICE curves ⇒ May **obscure heterogeneous effects** (interactions)

- **Example:** Feature x_1 = treatment dosage; x_2 = gender
 - ⇒ Males (increasing) and females (decreasing) respond differently to dosage
 - ⇒ PD curve (yellow) hides this divergence
- Plotting ICE and PD together helps detect interaction
- Diverse ICE shapes suggest interaction (but not with which feature)

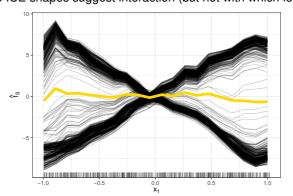




COMMENTS ON INTERACTIONS

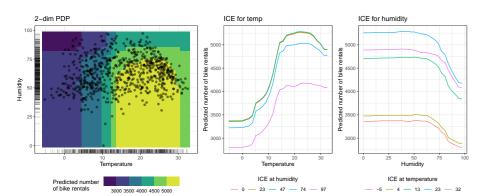
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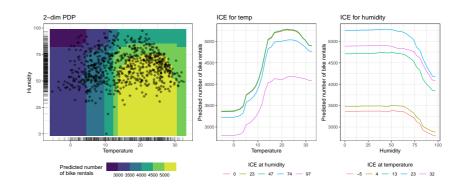
COMMENTS ON INTERACTIONS - 2D PD PLOT

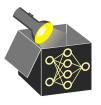




- Humidity and temperature interact at high values (see shape difference)
 - → Shape of ICE curves changes across different (higher) values of other feature
 - ICE (temp): At high humidity, temp effect flattens (pink line)
 - ICE (hum): At high temperature, humidity effect falls steeper (blue/pink)
- Most rentals occur at high temperature and low to medium humidity

COMMENTS ON INTERACTIONS - 2D PD PLOT





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CENTERED ICE PLOT (C-ICE) Goldstein et al. (2015)

Issue: Varying-intercept (stacked) ICE curves obscure shape heterogeneity **Solution:** Center ICE curves at fixed reference value, often $x' = \min(\mathbf{x}_S)$

⇒ Easier to identify heterogeneous shapes with c-ICE curves

$$\hat{t}_{S, \textit{cICE}}^{(i)}(\mathbf{x}_{S}) = \hat{t}(\mathbf{x}_{S}, \mathbf{x}_{-S}^{(i)}) - \hat{t}(x', \mathbf{x}_{-S}^{(i)}) = \hat{t}_{S}^{(i)}(\mathbf{x}_{S}) - \hat{t}_{S}^{(i)}(x')$$

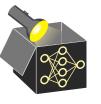


CENTERED ICE PLOT (C-ICE) • GOLDSTEIN_2015

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$$S_{S,CICE}(s) = (s, \xi_{-S}) - (x', \xi_{-S}) = S_{S}(s) - S_{S}(x')$$



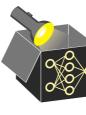
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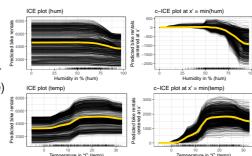
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Interpretation

- Yellow: c-PDP (mean of c-ICE)
- c-PDP: At 97% humidity, predicted rentals are 1000 fewer than at 0% humidity (on average)
- Opening of c-ICE curves: suggests interaction or varying effect across instances



CENTERED ICE PLOT (C-ICE) • GOLDSTEIN_2015

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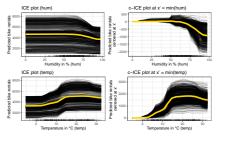
Solution: Center ICE curves at fixed reference value, often $x' = \min(s)$ \Rightarrow Easier to identify heterogeneous shapes with c-ICE curves

$$S_{SCICF}(s) = (s, \xi_{-S}) - (x', \xi_{-S}) = S_{S}(s) - S_{S}(x')$$



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CENTERED ICE PLOT (C-ICE)

Categorical features: c-ICE plots can be interpreted as in LMs due to reference value

Interpretation:

- The reference category is x' = SPRING
- Yellow crosses: Average rentals if we jump from SPRING to any other season
 ⇒ Number of bike rentals drops by
 ∼ 560 in WINTER and is slightly higher in SUMMER and FALL compared to SPRING

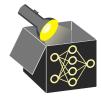


CENTERED ICE PLOT (C-ICE)

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Interpretation: • The referer

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