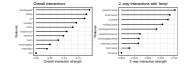
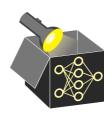
Interpretable Machine Learning

Friedman's H-Statistic



Learning goals

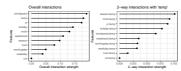
- Friedman's H-statistic with two purposes:
- Measure general *k*-way interactions between arbitrary features
- Measure a single feature's overall interaction strength



Interpretable Machine Learning

Functional Decompositions Friedman's H-Statistic



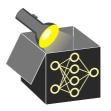


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2-way interaction:

• Two features j and k do not interact, if their 2-way interaction component in functional decomposition $g_{\{j,k\}}$ is 0



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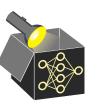
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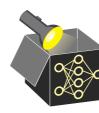
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This means: There are interactions
 ⇒ Every possible decomposition must contain some non-zero term g_{i,k} (x_i, x_k)



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Interpretable Machine Learning - 1 / 8



Again: remember GAMs

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3-way interaction:

• **Definition:** \hat{f} contains no 3-way interactions between features i, j, k, if corresponding 3-dimensional PD-function can be decomposed into lower-order terms:

$$\hat{f}_{\{ijk\},PD}(x_i,x_j,x_k) = g_\emptyset + g_i(x_i) + g_j(x_j) + g_k(x_k) + g_{\{i,i\}}(x_i,x_j) + g_{\{i,k\}}(x_i,x_k) + g_{\{i,j\}}(x_j,x_k)$$



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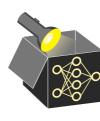


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• Example:

$$\hat{f}(x_1, x_2, x_3) = -2x_1 - 2\sin(x_3) + |x_1|x_2 - \sin(x_2x_3) + 1$$



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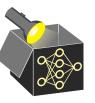
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• Note: Again using centered PD-functions $\hat{f}_{S,PD}^c$ instead of components g_S \leadsto things get complicated, e.g. for 3 features, definition becomes:

 $\hat{f}_{\{jik\},PD}(x_i,x_i,x_k) = g_{\emptyset} + g_i(x_i) + g_i(x_i) + g_k(x_k)$

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k-way interaction:

• **Analogous** for general k-way interactions between features $S = \{i_1, i_2, \dots, i_k\}$: No k-way interaction, if

$$\hat{f}_{\mathcal{S},\mathsf{PD}}(x_{i_1},x_{i_2},\ldots,x_{i_k}) = \sum_{\substack{V \subseteq \mathcal{S} \ |V| < k}} g_V(\mathbf{x}_V) = \sum_{\substack{V \subseteq \mathcal{S} \ |V| < k}} g_V(\mathbf{x}_V)$$

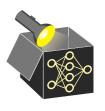


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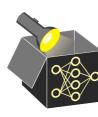


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- Question: Does feature *j* interact with any other feature at all?
- \Rightarrow H-statistic analogous to 2-way interactions, but for feature sets $S = \{j\}$ and $-S = \{1, \dots, p\} \setminus \{j\}$ instead of two single features:



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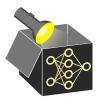
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Overall interaction:

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- -j denotes -S = {1,...,p} \ {j}, i.e. all other features
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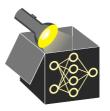
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2-WAY INTERACTION STRENGTH

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Interpretable Machine Learning - 4 / 8

- Question: How to measure interaction strength without computing functional decomposition components g_S?
- Idea: Only use centered PD-functions

$$\hat{f}^{c}_{\{jk\},PD}(x_{j},x_{k}) = \hat{f}^{c}_{j,PD}(x_{j}) + \hat{f}^{c}_{k,PD}(x_{k})$$
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Interpretable Machine Learning - 4 / 8

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• **H-statistic** for 2-way interaction between feature *j* and *k*:

$$H_{jk}^{2} = \frac{\operatorname{Var}\left[\hat{f}_{jk,PD}^{c}(X_{j},X_{k}) - \hat{f}_{j,PD}^{c}(X_{j}) - \hat{f}_{k,PD}^{c}(X_{k})\right]}{\operatorname{Var}\left[\hat{f}_{jk,PD}^{c}(X_{j},X_{k})\right]}$$

$$= \frac{\sum_{i=1}^{n}\left(\hat{f}_{jk,PD}^{c}(x_{j}^{(i)},x_{k}^{(i)}) - \hat{f}_{j,PD}^{c}(x_{j}^{(i)}) - \hat{f}_{k,PD}^{c}(x_{k}^{(i)})\right)^{2}}{\sum_{i=1}^{n}\left(\hat{f}_{jk,PD}^{c}(x_{j}^{(i)},x_{k}^{(i)})\right)^{2}}$$



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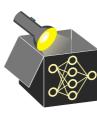
$$\hat{f}^{c}_{\{jk\},PD}(x_{j},x_{k}) = \hat{f}^{c}_{j,PD}(x_{j}) + \hat{f}^{c}_{k,PD}(x_{k})$$
?

• **H-statistic** for 2-way interaction between feature *j* and *k*:

$$H_{jk}^{2} = \frac{\operatorname{Var}\left[\hat{f}_{jk,PD}^{c}(X_{j},X_{k}) - \hat{f}_{j,PD}^{c}(X_{j}) - \hat{f}_{k,PD}^{c}(X_{k})\right]}{\operatorname{Var}\left[\hat{f}_{jk,PD}^{c}(X_{j},X_{k})\right]}$$

$$= \frac{\sum_{i=1}^{n} \left(\hat{f}_{jk,PD}^{c}(x_{j}^{(i)},x_{k}^{(i)}) - \hat{f}_{j,PD}^{c}(x_{j}^{(i)}) - \hat{f}_{k,PD}^{c}(x_{k}^{(i)})\right)^{2}}{\sum_{i=1}^{n} \left(\hat{f}_{jk,PD}^{c}(x_{j}^{(i)},x_{k}^{(i)})\right)^{2}}$$

 \Rightarrow H_{jk}^2 measures strength of this interaction quantitatively H_{ik}^2 small (close to 0) for weak interaction, close to 1 for strong interaction



2-WAY INTERACTION STRENGTH

- **Question:** How to measure interaction strength without computing functional decomposition components g_S ?
- Idea: Only use centered PD-functions

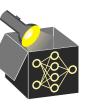
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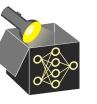
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Note: Again, definition also usable without probabilities or data distrib.



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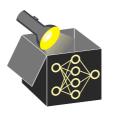
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H-STATISTIC: EXAMPLES

Note: Again, definition also usable without any probability or data distribution

Example

$$\begin{aligned} \hat{f}(x_1, x_2) &= 4 - 2x_1 + 0.3e^{x_2} + |x_1|x_2 \quad (x_1, x_2) \in [0, 1]^2 \\ \hat{f}_{1, PD}^c(x_1) &= -2x_1 + 0.5|x_1| + 0.75 \\ \hat{f}_{2, PD}^c(x_2) &= 0.3e^{x_2} + 0.5x_2 - 0.3e + 0.05 \\ \hat{f}_{1, 2; PD}^c(x_1, x_2) &= 1.05 - 2x_1 + 0.3e^{x_2} + |x_1|x_2 - 0.3e \end{aligned}$$



H-STATISTIC: EXAMPLES

Note: Again, definition also usable without any probability or data distribution **Example**



$$\hat{f}(x_1, x_2) = 4 - 2x_1 + 0.3e^{x_2} + |x_1|x_2 \quad (x_1, x_2) \in [0, 1]^2$$

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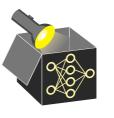
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H-STATISTIC: EXAMPLES

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$$\begin{split} \hat{f}(x_1, x_2) &= 4 - 2x_1 + 0.3e^{x_2} + |x_1|x_2 \quad (x_1, x_2) \in [0, 1]^2 \\ \hat{f}_{1, PD}^c(x_1) &= -2x_1 + 0.5|x_1| + 0.75 \\ \hat{f}_{2, PD}^c(x_2) &= 0.3e^{x_2} + 0.5x_2 - 0.3e + 0.05 \\ \hat{f}_{1, 2; PD}^c(x_1, x_2) &= 1.05 - 2x_1 + 0.3e^{x_2} + |x_1|x_2 - 0.3e \\ &\Longrightarrow H_{12}^2 = \frac{\text{Var}\left[\hat{f}_{jk, PD}^c(X_j, X_k) - \hat{f}_{j, PD}^c(X_j) - \hat{f}_{k, PD}^c(X_k)\right]}{\text{Var}\left[\hat{f}_{jk, PD}^c(X_j, X_k)\right]} \\ &= \frac{\mathbb{E}\left[\left(|x_1|x_2 - 0.5|x_1| - 0.5x_2 + 0.25\right)^2\right]}{\mathbb{E}\left[\left(1.05 - 2x_1 + 0.3e^{x_2} + |x_1|x_2 - 0.3e\right)^2\right]} > 0 \end{split}$$



H-STATISTIC: EXAMPLES

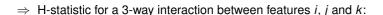
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• Same idea as for 2-way, but different formula (see before):

$$\hat{f}_{\{ijk\},PD}^{c}(x_i,x_j,x_k) = \hat{f}_{\{ij\},PD}^{c}(x_i,x_j) + \hat{f}_{\{ik\},PD}^{c}(x_i,x_k) + \hat{f}_{\{jk\},PD}^{c}(x_j,x_k)
- \hat{f}_{i,PD}^{c}(x_i) - \hat{f}_{i,PD}^{c}(x_j) - \hat{f}_{k,PD}^{c}(x_k)$$



$$H_{ijk}^{2} = \frac{\mathsf{Var}\left[\hat{f}_{ijk,PD}^{c}(X_{i},X_{j},X_{k}) - \hat{f}_{ij,PD}^{c}(X_{i},X_{j}) - \hat{f}_{ik,PD}^{c}(X_{i},X_{k}) - \hat{f}_{jk,PD}^{c}(X_{j},X_{k})\right]}{+\hat{f}_{i,PD}^{c}(X_{i}) + \hat{f}_{i,PD}^{c}(X_{j}) + \hat{f}_{k,PD}^{c}(X_{k})}}{\mathsf{Var}\left[\hat{f}_{ijk,PD}^{c}(X_{i},X_{j},X_{k})\right]}$$

Analogous for higher order interactions, but more complicated



3-WAY INTERACTION STRENGTH

• Same idea as for 2-way, but different formula (see before):

$$\hat{f}_{\{ijk\},PD}^{c}(x_i, x_j, x_k) = \hat{f}_{\{ij\},PD}^{c}(x_i, x_j) + \hat{f}_{\{ik\},PD}^{c}(x_i, x_k) + \hat{f}_{\{jk\},PD}^{c}(x_j, x_k)
- \hat{f}_{i,PD}^{c}(x_i) - \hat{f}_{i,PD}^{c}(x_j) - \hat{f}_{k,PD}^{c}(x_k)$$

0000

 \Rightarrow H-statistic for a 3-way interaction between features *i*, *j* and *k*:

$$H_{ijk}^{2} = \frac{\text{Var}\left[\hat{f}_{ijk,PD}^{c}(X_{i},X_{j},X_{k}) - \hat{f}_{ij,PD}^{c}(X_{i},X_{j}) - \hat{f}_{ik,PD}^{c}(X_{i},X_{k}) - \hat{f}_{jk,PD}^{c}(X_{j},X_{k})\right]}{+\hat{f}_{i,PD}^{c}(X_{i}) + \hat{f}_{j,PD}^{c}(X_{j}) + \hat{f}_{k,PD}^{c}(X_{k})}}{\text{Var}\left[\hat{f}_{ijk,PD}^{c}(X_{i},X_{j},X_{k})\right]}$$

Analogous for higher order interactions, but more complicated

Interpretable Machine Learning - 6 / 8

OVERALL INTERACTION STRENGTH

- Measure overall strength of interactions between feature *j* and all other features
- ⇒ H-statistic analogous to 2-way interaction:

$$H_{j}^{2} = \frac{\text{Var}\left[\hat{f}^{c}(\mathbf{X}) - \hat{f}_{j,PD}^{c}(X_{j}) - \hat{f}_{-j,PD}^{c}(\mathbf{X}_{-j})\right]}{\text{Var}\left[\hat{f}^{c}(\mathbf{X})\right]}$$

$$= \frac{\sum_{i=1}^{n} \left(\hat{f}^{c}(\mathbf{x}^{(i)}) - \hat{f}_{j,PD}^{c}(x_{j}^{(i)}) - \hat{f}_{-j,PD}^{c}(\mathbf{x}_{-j}^{(i)})\right)^{2}}{\sum_{i=1}^{n} \left(\hat{f}^{c}(\mathbf{x}^{(i)})\right)^{2}}$$



OVERALL INTERACTION STRENGTH

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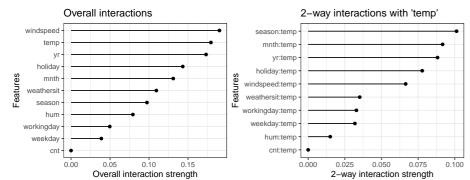
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Interpretable Machine Learning - 7/8

H-STATISTIC: EXAMPLE

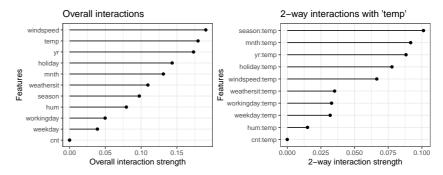
Measure interactions of a random forest for the bike data set





H-STATISTIC: EXAMPLE

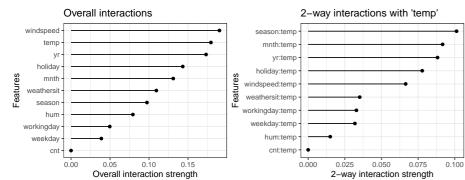
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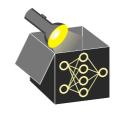




H-STATISTIC: EXAMPLE

Measure interactions of a random forest for the bike data set



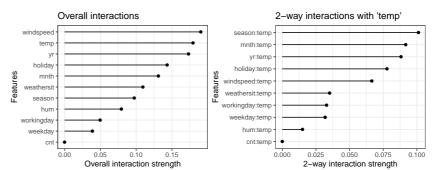


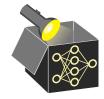
Remarks and Conclusion:

- H-statistic provides general definition of interactions + an algorithm for computation
 - Also adjustable to categorical / discrete features and / or function values
- For interaction order k still needs $\approx 2^k$ PD-functions
- Statistical test for whether interactions are present using this statistic

H-STATISTIC: EXAMPLE

Measure interactions of a random forest for the bike data set





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Interpretable Machine Learning - 8 / 8