

Feature importance

Bram Droppers

Workshop

- Three sections
 - Impurity feature importance
 - Permutation feature importance
 - SHAP feature importance



Workshop

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 - Impurity feature importance
 - Permutation feature importance
 - SHAP feature importance
- Per section
 - Small exercise
 - Presentation and questions



Why feature importance?



Why feature importance?

- Common sense check
- Uncertainty analysis
- Reducing model size and complexity



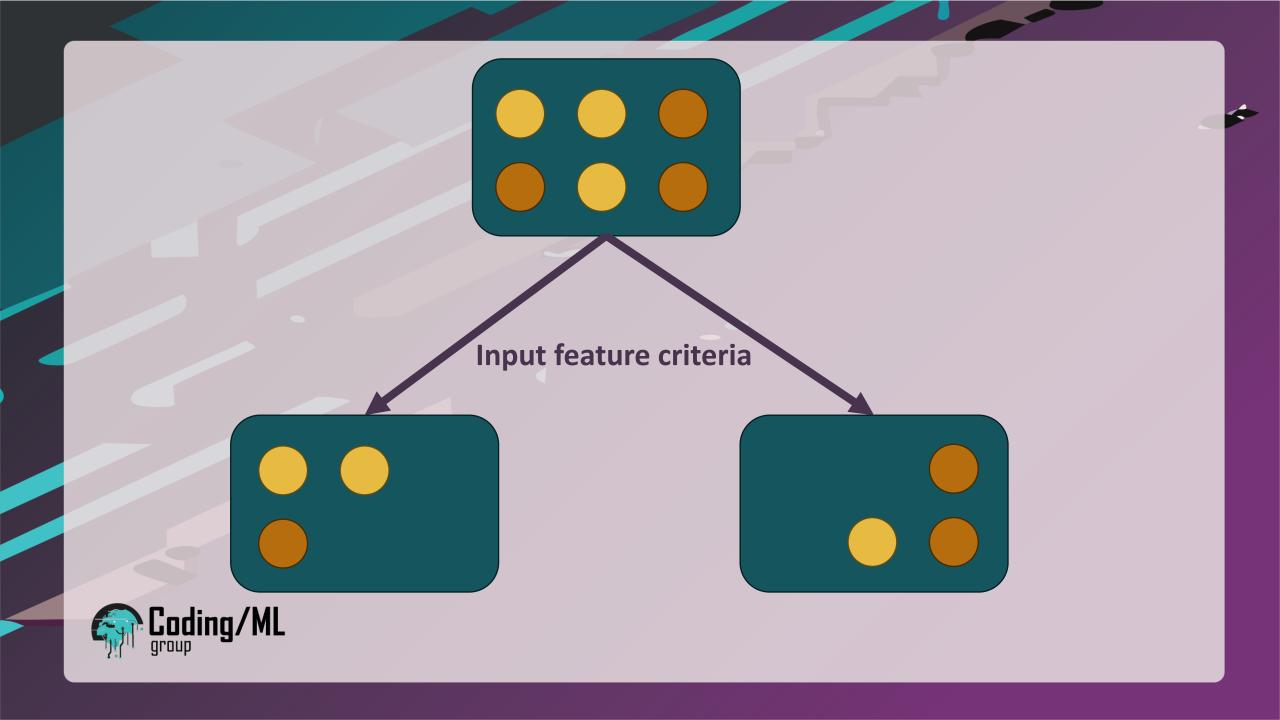
Impurity feature importance



Impurity feature importance

- Also called:
 - Gini importance
 - Mean decrease impurity
- Decrease in node impurity, weighted by the probability of reaching that node





Impurity feature importance

- Also called:
 - Gini importance
 - Mean decrease impurity
- Decrease in node impurity, weighted by the probability of reaching that node
- + Already calculated

- Only for random-forest models



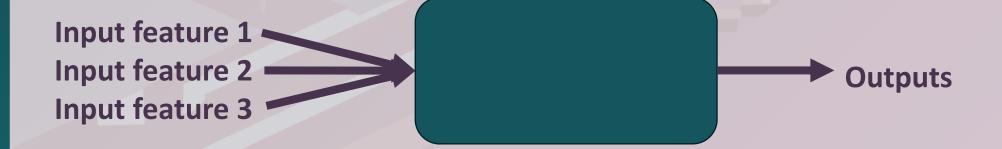
Permutation feature importance



Permutation feature importance

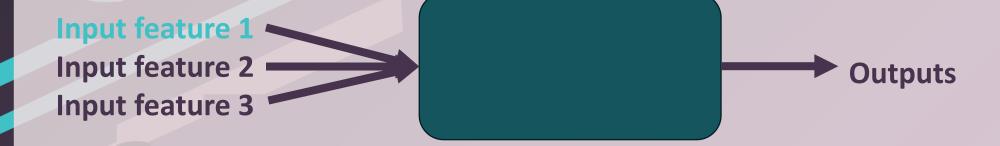
• Difference in model output after permutation of input features



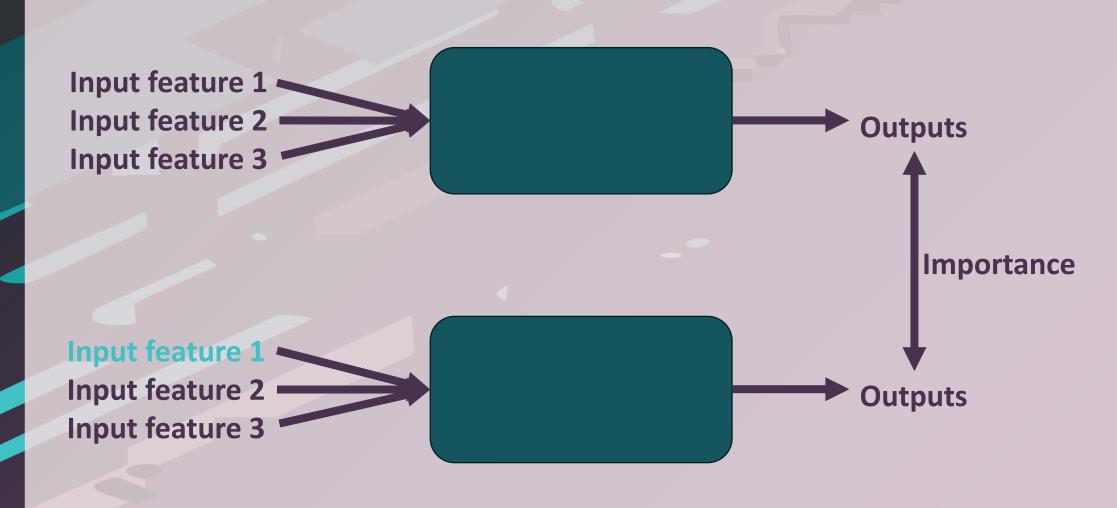








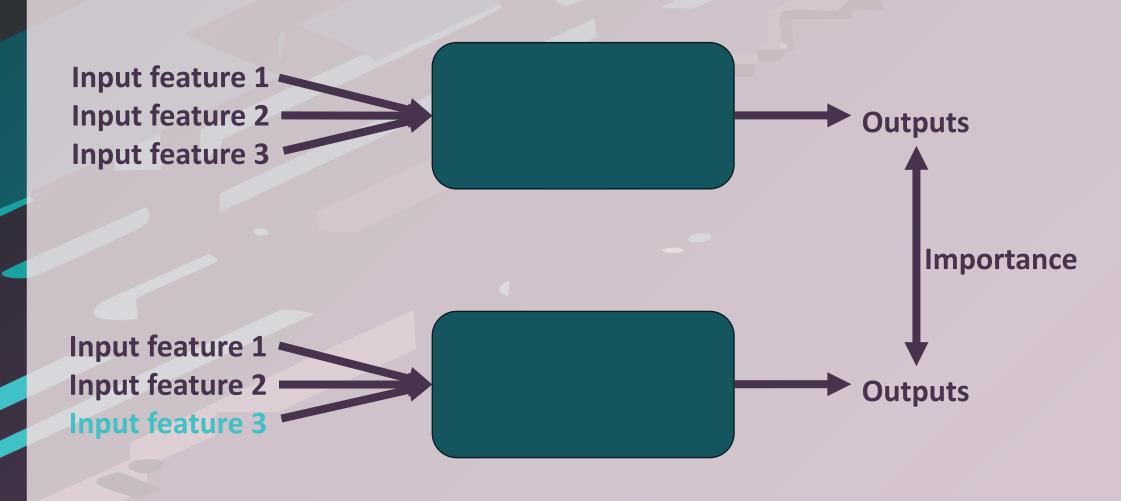














Permutation feature importance

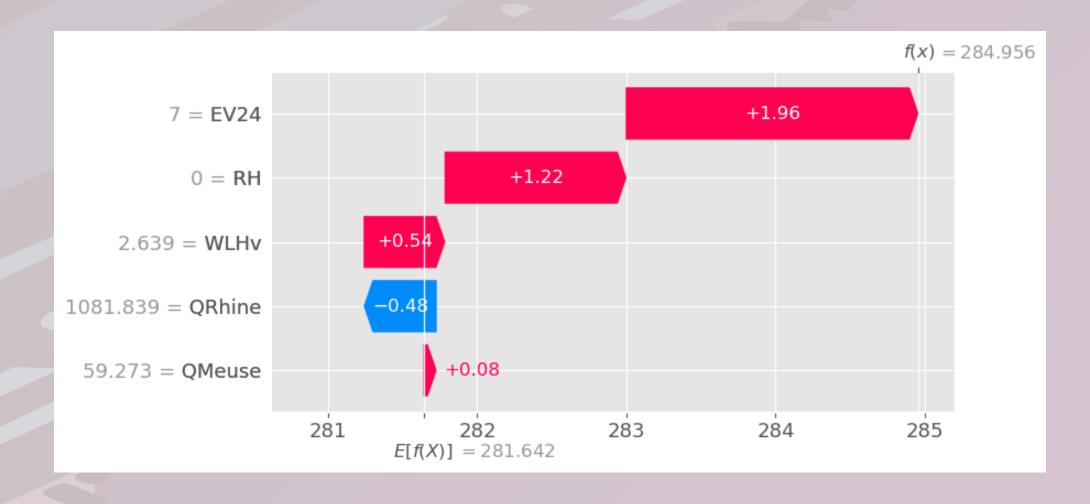
- Difference in model output after permutation of input features
- + Applicable to all models
- Slow
- Limited accounting for complex non-linear interactions





 Represents the marginal contribution of a feature's value to the prediction averaged over all possible combinations





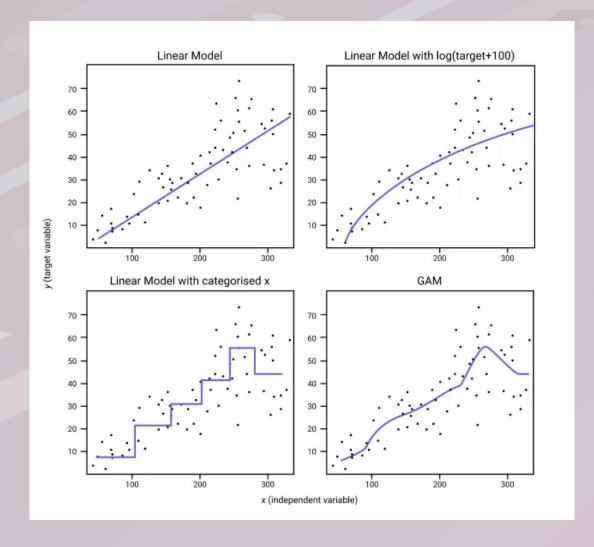


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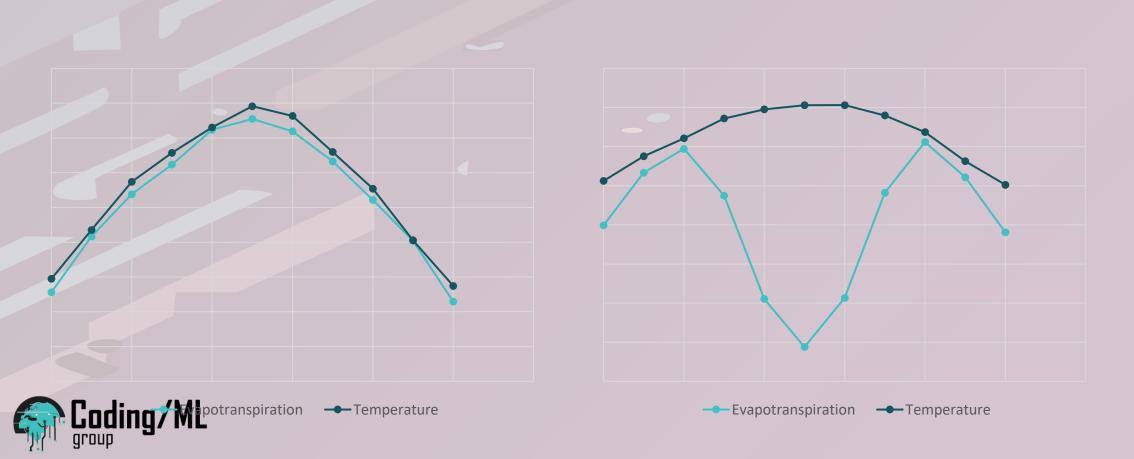
- Even slower





- We do not know how our models handle correlated input features
 - Ignore one
 - Use both





- We do not know how our models handle correlated input features
 - Ignore one
 - Use both
- This is reflected in the feature importance analysis



- Combine correlated features
- Omit correlated features
- Design a better train-test set





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