

Binning in H2O Decision Trees

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 High Cardinality Features
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 Bin the levels in a categorical column according to "phins, cat" parameter.
 - Bin the levels in a categorical column according to "nbins_cat" parameter.
 More Bins, More Likely To Overfit Increasing the number of bins can lead to splits on a single category, which can lead to overfitting.

Supervised Learning:

TREE ENSEMBLES

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