

Yandex



Dealing with categorical features and overfitting

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Categorical features

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Image: www.petsworld.in

One-hot encoding

[proton, pion, kaon] \rightarrow $[[1, 0, 0], [0, 1, 0], [0, 0, 1]]$

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› Doesn't scale well with the number of categories

CTR (aka click-through ratio)

For each pair
(target_class, categorical_feature_value):

$$\text{ctr}_i = \frac{\text{countInClass} + \text{prior}}{\text{totalCount} + 1}$$

- › countInClass – number of objects in the i -th class with the current categorical feature value
- › prior – algorithm parameter
- › totalCount – total number of objects with the current categorical feature value

CTR example

fruit	target	ctr
apple	0	0.625
orange	0	0.25
apple	1	0.625
apple	1	0.625

prior = 0.5

Classes counter

For each pair

(target_class, categorical_feature_value):

$$\text{count}_i = \frac{\text{curCount} + \text{prior}}{\text{totalCount} + 1}$$

- › curCount – number of objects with the current categorical feature value
- › prior – algorithm parameter
- › totalCount – total number of objects

Counters example

fruit	target	ctr	counter
apple	0	0.625	0.7
orange	0	0.25	0.3
apple	1	0.625	0.7
apple	1	0.625	0.7

prior = 0.5

Overfitting

Gradients bias in gradient boosting

- › Each subsequent tree is fit to the gradient between the current predictions on train and the true labels

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- › Each subsequent tree is fit to the gradient between the current predictions on train and the true labels
- › The gradient is estimated using the model fitted on the very dataset used for training
- › The gradients are likely to be overfitted

Dynamic boosting



- › Order data randomly

Dynamic boosting



- › Order data randomly
- › For each element maintain prediction based on the previous model elements

GB matters

Gradient boosting is the workhorse of Yandex Internet business

Meet CatBoost



- › Gradient boosting on decision trees
- › Categorical features handling (even more advanced than discussed!)
- › A novel boosting scheme (submitted to NIPS)
- › Released into open source by Yandex on Tuesday
- › Used in the LHCb PID

Contacts

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