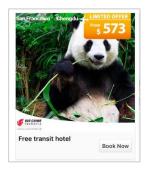
Click-Through-Rate Prediction Scalable Factorization Machines

Binjie Lai, Miao Lu, Wanyan Xie MSAN, USF



Click-Through-Rate Prediction







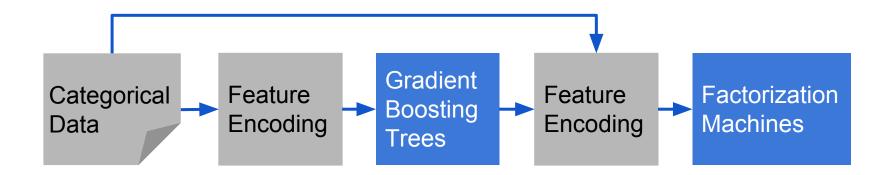


Kaggle Dataset : Avazu CTR Prediction

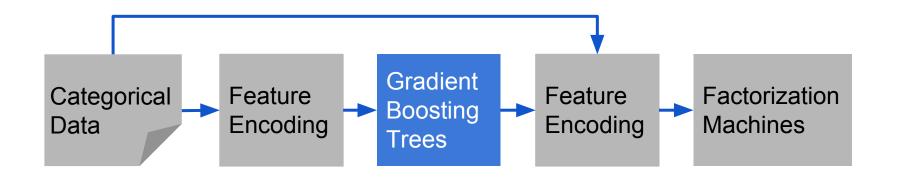
Train: **5.88 GB** Test: **629 MB**

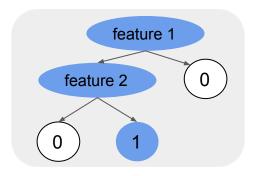
Click	Banner _position	App _id	App _category	Device _id	Device _model		Impressions
0	0	ecad2386	07d7df22	a99f214a	44956a24		
1	0	ecad2386	07d7df22	a99f214a	44956a24		~ 36 million
0	2	7e091613	f028772b	6afc734f	e352da7		
?	1	85f751fd	c4e18dd6	6afc734f	e352da7		~ 4 million
	\		_\ /			'	.,
label			23 categori features	cal			

Data Source: https://www.kaggle.com/c/avazu-ctr-prediction

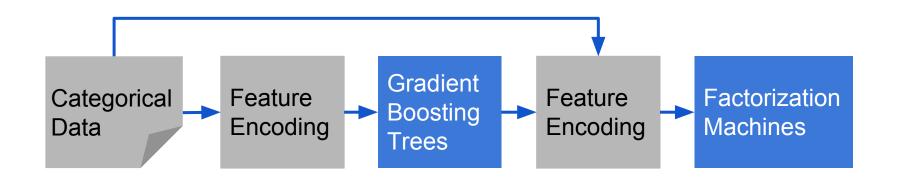


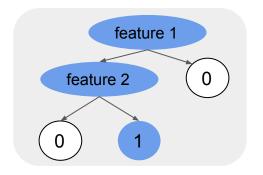


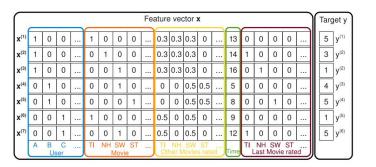






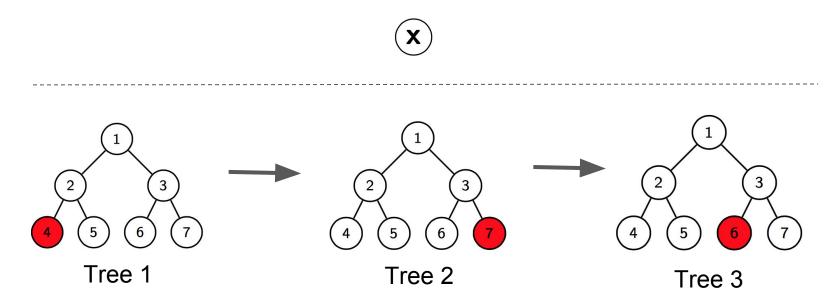








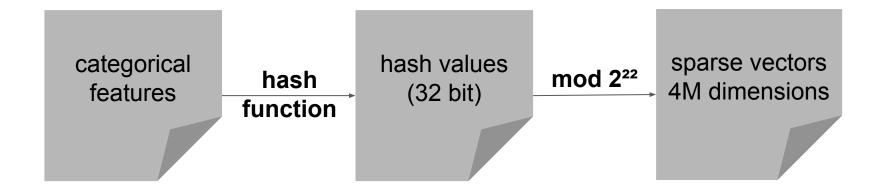
Feature Engineering: Gradient Boosting Trees Encoding



new features: tree1: 4, tree 2: 7, tree 3: 6



Feature Encoding





Factorization Machines

logistic regression:

$$y(x) = \sigma(\omega_0 + \sum_{i=1}^{n} \omega_i x_i + \sum_{i=1}^{n} \sum_{j=i+1}^{n} w_{i,j} x_i x_j)$$



Factorization Machines

logistic regression:

$$y(x) = \sigma(\omega_0 + \sum_{i=1}^n \omega_i x_i + \sum_{i=1}^n \sum_{j=i+1}^n w_{i,j} x_i x_j)$$

$$\mathbf{w}_{i:4\mathbf{M}\times 4\mathbf{M}}$$



Factorization Machines

logistic regression:

$$y(x) = \sigma(\omega_0 + \sum_{i=1}^n \omega_i x_i + \sum_{i=1}^n \sum_{j=i+1}^n w_{i,j} x_i x_j)$$
 $w_{i,j} = < v_i, v_j >$
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factorization machines:

$$y(x) = \sigma(\omega_0 + \sum_{i=1}^n \omega_i x_i + \sum_{i=1}^n \sum_{j=i+1}^n \langle v_i, v_j \rangle x_i x_j)$$



Factorization Machines Optimization

Markov Chain Monte Carlo

-- Sampling model parameters

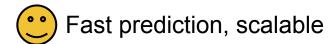


Prediction in training time

Built on a single machine

Stocastic Gradient Desent (SGD)

-- Minimizing loss function



Lots of hyperparameters

Built on Spark
Distributed Mini-Batch SGD



Method	AUC	Log loss
FM + GBT	0.7456	0.3933
FM	0.7431	0.3952
LR + GBT	0.7432	0.3945
LR	0.7408	0.4006

FM: Factorization Machines with Markov Chain Monte Carlo

GBT: Gradient Boosting Trees



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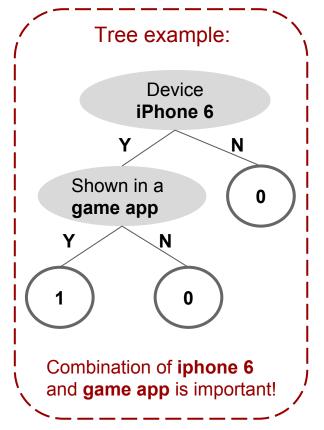
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Running Time

Step	Single Machine
Encoding	~ 5 hrs
GBT	~ 55 hrs
FM / LR	~ 2 hrs

Single machine: EC2 with 160 GB memory

Distributed System: Spark EC2 with 10 nodes



Running Time

Step	Single Machine	Distributed System
Encoding	~ 5 hrs	~ 40 min
GBT	~ 55 hrs	~ 6 hrs
FM / LR	~ 2 hrs	~ 30 min

Single machine: EC2 with 160 GB memory
Distributed System: Spark EC2 with 10 nodes



Summary

- Automatic feature engineeing
- Capturing high order interactions under sparsity
- Scalable algorithm
- Potential for production





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