Predict future sales

Kaggle competition https://www.kaggle.com/c/competitive-data-science-predict-future-sales/data

Goal

- Predict total number of sales for every product and store in the next month
 - given daily sales data provided by company 1C Company
- Time series dataset, time period-33 months

In [11]: train.shape, test.shape
Out[11]: ((2935849, 6), (214200, 3))

[n [5]:	train.head()											
Out[5]:		date	date_block_num	shop_id	item_id	item_price	item_cnt_day					
	0	02.01.2013	0	59	22154	999.0	1.0					
	1	03.01.2013	0	25	2552	899.0	1.0					
	2	05.01.2013	0	25	2552	899.0	-1.0					
	3	06.01.2013	0	25	2554	1709.0	1.0					
	4	15.01.2013	0	25	2555	1099.0	1.0					

<pre>In [6]: test.head()</pre>										
	ID	shop_id	item_id							
0	0	5	<mark>5037</mark>							
1	1	5	5320							
2	2	5	<mark>5233</mark>							
3	3	5	5232							
4	4	5	5268							
	te 0 1 2 3 4	ID 0 0 1 1 2 2 3 3 4 4	test.head() ID shop_id 0 0 5 1 1 5 2 2 5 3 3 5 4 4 5							

Data

Sales



Items



Categories



Data exploration

• Sales by month



Data exploration

- Product marketability
- 2371 products that have been sold once



freq	item name
31340	Corporate package white shirt 1C Interest (34
9408	Playstation Store replenishment wallet: Map pa
9067	Receiving cash for 1C-line
7479	Diablo III [PC, Jewel, Russian version]
6853	Kaspersky Internet Security Multi-Device Russi

Number 21807	of un	ique pr	odu	cts:	22170
	freq	number	of	pro	ducts
Θ	1			5	2371
1	2				1054
2	3				669
3	4				540
4	5				470
1251	6853				1
1252	7479				1
1253	9067				1
1254	9408				1
1255	31340				1

[1256 rows x 2 columns]

Preprocessing

- 1. Daily sale records -> monthly sale records
 - we are supposed to predict the total number of product sales in next month
- 2. Removing outliers
- 3. Negative sales (product returns)
- 4. Constructing additional features
- 5. Validation set



	ID	shop_id	item_id	0	1	2	3	4	5	6	 24	25	26	27	28	29	30	31	32	33
0	0	5	5037	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 2.0	0.0	0.0	0.0	1.0	1.0	1.0	3.0	1.0	0.0
1	1	5	5320	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
2	2	5	5233	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0	3.0	2.0	0.0	1.0	3.0	1.0
3	3	5	5232	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0	0.0	0.0
4	4	5	5268	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
								112			 									
214195	<mark>21419</mark> 5	45	18454	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 2.0	1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	1.0
214196	214196	45	16188	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
214197	214197	45	15757	1.0	0.0	0.0	0.0	0.0	0.0	0.0	 1.0	1.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
214198	<mark>214198</mark>	45	19648	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
214199	214199	45	969	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0

Results

- Models
 - LGBMRegressor
 - **0.88**
 - XGBRegressor
 - **1**.02
 - RNN (PyTorch)
 - 0.83
- Score Root mean squared error

#	Team Name	Notebook	Team Members	Score 🕝	Entries	Last
1	KDJ2020		(9)	0.75368	376	8mo
2	Shorokhov Sergey			0.76955	172	2Y
3	VNPT@DS		🌚 🤿 🕥	0.78399	148	1y
4	Konstantin Yakovlev			0.79215	210	ЗY
5	b_b		9	0.79358	195	ЗY

Feature engineering

- Lag features
- Delta_price_lag



Issue - missing training samples

- We have no sale history in the train set for half of the products that we have to predict
- We can try to generate some representative values
 - Cannot validate these values

Unique (product, shop) pairs in train/test



Solution?

- Idea
 - For every pair <item_id, shop_id> not included in train set, do:
 - 1. Mean of all pairs <item_id, X>
 - If there are no pairs <item_id, X> in train set (exactly 15246 cases), do:
 - Mean of all pairs <Y, shop_id>
 - In case there are no pairs <Y, shop_id> (0 cases) in train set set the value to zero
 - 3. Use this value as a prediction

- Did it work? Not really
- Worse results

Things worth trying

- CatBoost
- AutoML
- Additional feature extraction
- Still plenty of time to submit: Competition ends in Jan 1. 2023

Thanks for your attention