

Predicting sales prices for US houses

<u>Team 1</u>

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BI meets Data Science

- Learn PowerBI
- Explore use of R Studio (and similar)
- Experiment with Power BI + R
- Create a working ML pipeline
- Sharpen/update data science skills
- Have fun 🙂



Image source:

https://caiomsouza.medium.com/difference-between-bi-business-intelligence-anddata-science-1a9c7628bbdb (23-09-2022)

Starting with some science before diving into the data (swamp)

- Variety of approaches:
 - Machine learning (Komagome-Towne, 2016; Ravikumar, 2017; Phan, 2018; Truong et al., 2020)
 - Deep learning (Wang et al., 2021)
 - Lineair- and Logistic regression (He, He, 2021), Fuzzy (Sarip, 2016) and Multiple regression (Zhang, 2021)
 - Time series (Wang, Juntao, et al. 2018)
 - AutoML (Li et al., 2020)
 - Particle swarm optimization (Zhou, 2021)
- Different sources:
 - Mainly based on historical data
 - Economic parameters (Li and Chu, 2017)
 - News data (Kirkeby and Larsen, 2021)

Clearly no "silver bullet" solution or generalized set of features

Questions for this week:

- How to test historical, spatial and temporal influences?
- Which risk of bias is in the data?

Questions for later/others:

- Which indices correlate with housing prices?
- What influence does public opinion and sentiment have?



Use of PowerBl Data crunching

- Dataset merge
- Data cleaning
- Data exploration/visualizations
- Correlation map
- Influencing factors
- Experimental AI: Q&A
- Experimental: integrate R script

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Use of PowerBI Correlations and AI-based Q&A



(i) Help Q&A understand people better by adding synonyms. Add synonyms now	\sim
💭 what is the average log sales price of neighborhood BrDale (neighborhood)	명 🕸
	(i)

11,55 Average of LogSalePrice

Use of R in Power BI

- Data visualization (library: corrplot)
- R script editor in PowerBI
- Explored functionalities in R
- Continued in RapidMiner

R script editor									
1	#	The following code to create a dataf	Fran						
2									
3	#	<pre>dataset <- data.frame(LogSalePrice)</pre>							
4	#	dataset <- unique(dataset)							



Use of RapidMiner

- Data preparation
- Statistics/visualizations
- Outlier detection
- Modeling with AutoML (6 models + correlation map)
- Model evaluation and selection: Gradient Boosted Trees
- Model deployment
- Scoring



Predicting future sales prices



Important Factors for Prediction

Model	Relative Error	Standard Deviation	Total Time	Training Time (1,000 Rows)	Scoring Time (1,000 Rows)
Generalized Linear Model	0,70%	0,08%	1369,0	42,2	6,9
Deep Learning	0,77%	0,07%	5162,0	923,9	133,2
Decision Tree	1,40%	0,09%	1331,0	11,8	12,1
Random Forest	0,86%	0,12%	21752,0	204,8	192,0
Gradient Boosted Trees	0,69%	0,04%	32132,0	341,9	57,1
Support Vector Machine	138,15%	0,34%	27709,0	3815,2	207,6

Concluding remarks

- Nice combination of BI & Data Science
- Quite a lot features in PowerBI that go beyond BI
- RapidMiner really speeds up process with TurboPrep and AutoML - and also offers a variety of statistics / visualizations
- Working ML pipeline in 1 week based on Gradient Boosted Trees and iterative development

Didn't go in-depth with:

- R scripting
- (Hyper)parameter tuning

PowerBI correlations limited to numeric values

We had fun and got predictions :)

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