# Kaggle Sprint: Real-Estate Price Estimation

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# **Speakers**

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KTU PhD student in operation research in logistics specializing in agentbased modelling and machine learning.

#### Andrius Kriščiūnas

KTU Informatics faculty lecturer & researcher specializing in optimization and approximation techniques.

Workshop focuses on basic approach and not state of the art solutions, we focus on research not production. Thus, more advanced techniques to place in TOP Kaggle places will not be presented.

# Agenda

- Introduction to real estate price estimation
- Understanding the algorithms
- Exploratory analysis of the dataset
- Features engineering
- Main algorithms usage: XGBoost, LightGBM
- Result comparison
- Useful resources
- Questions

## Goal

- How to decide for how much to sell a real estate object?
- How to evaluate tax for real estate?
- What size of a loan and what size of interests can be provided?
- Etc.

Usually expert evaluation and object comparison are used to answer the questions, and not data science approaches.

### Real estate evaluation

Real estate evaluations being used by investors, government, banks, citizens:

- To plan their households;
- To plan infrastructure projects;
- To plan bank loans and interest schemes;
- Etc.

The current methodology of real estate price evaluation usually uses expert evaluation, because:

- It is difficult to integrate multiple sources of data
- The majority of experts does not comprehends mathematical evaluation methods

## Data sources to improve estimation

- Real estate demand <a href="https://www.aruodas.lt/">https://www.aruodas.lt/</a>
- Climate <u>www.windguru.com</u>
- Comodity prices <u>www.indexmundi.com</u>
- Natural resources <u>www.usgs.gov</u>
- Macro indicators <a href="https://www.worldbank.org/">https://www.oecd.org/</a>, <a href="https://www.oecd.org/">https://www.oecd.org/</a>, <a href="https://www.oecd.org/">https://www.oecd.org/</a>, <a href="https://data.un.org/">https://data.un.org/</a>
- Ortophotograpgy <u>www.copernicus.eu</u>
- Transport infrastructure statistics <a href="https://github.com/graphhopper/open-traffic-collection">https://github.com/graphhopper/open-traffic-collection</a>
- Assocications <a href="https://www.fefac.eu/">https://www.fefac.eu/</a>
- Import, Export <a href="https://comtrade.un.org/">https://comtrade.un.org/</a>
- Google trends <a href="https://trends.google.com">https://trends.google.com</a>
- Stock prices www.nasdaq.com
- Unstructured data (TF-IDF) reports, news

## Main issue

#### Different methodologies and revisions:

- The Global Industry Classification Standard (GICS)
- Classification of Economic Activities (EVRK)
- Harmonized Commodity Description and Coding Systems
- Common classification of territorial units for statistics (NUTS)
- And so on

## Data archives for research & learning

- UCI Machine Learning Repository <a href="https://archive.ics.uci.edu">https://archive.ics.uci.edu</a>
- Kaggle <u>www.kaggle.com</u>
- Other?

## Which data can you use?

Which data you have before making the decision?

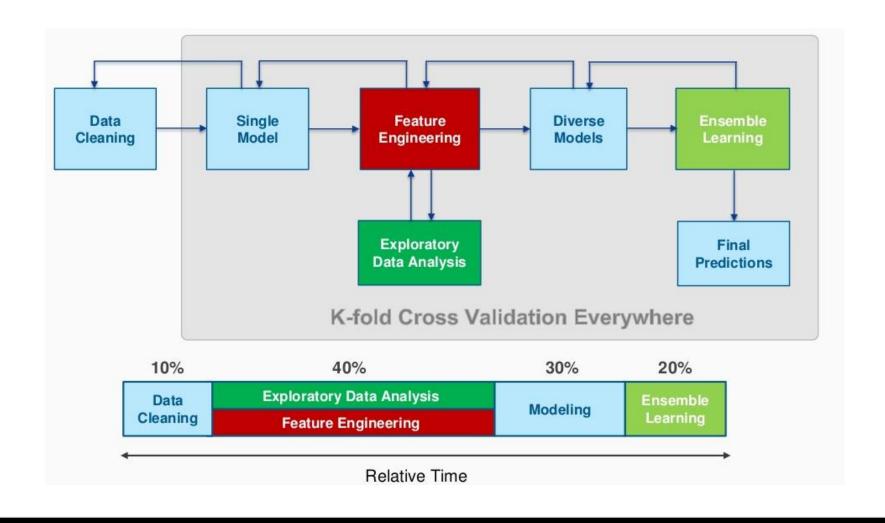


Source: <a href="https://www.oreilly.com/ideas/drivetrain-approach-data-products">https://www.oreilly.com/ideas/drivetrain-approach-data-products</a>

# Kaggle competition

- Kaggle is an online community of data scientists and machine learners, that provides competitions to win prizes and develop algorithms for companies.
- Kaggle is a great tool to learn machine learning and have a benchmarking to compare your model.
- Data source used: Sberbank Russian Housing Market

## **ML** implementation process



# INTRODUCTION TO MATHEMATICS BEHIND THE ML ALGORITHMS

PhD, Andrius Kriščiūnas

## Libraries

- 1. Run anaconda prompt
- 2. cd to directory
- 3. pip install -r requirements.txt

- numpy
- panda
- matplotlib
- seaborn
- sklearn
- xgboost
- lightgbm
- datetime
- shap
- pickle

# **Exploratory analysis of the dataset**

- Dependencies of target variables and additional variables
- Distribution of data
- Dependencies between target variable and additional variables
- Missing values
- Outlier detection
- Data mistake fixing

## Features engineering

- Transform data to improve dependency, scale data, fit closer to normal distribution (e.g. log, 1/x etc.)
- Identify main variable influence to target variable (e.g. correlation analysis)
- Remove similar variables (multicollinearity)
- Merge similar variables (e.g. Principal component analysis)
- Add other variables based on experience and creativity

## Work automation

#### **Pipeline**

Sequentially apply a list of transforms and a final estimator. Intermediate steps of the pipeline must be 'transforms', that is, they must implement fit and transform methods.

E.g. Sklearn pipeline

# XGBoost and LightGBM

- **XGBoost** is an optimized distributed gradient boosting library designed to be highly efficient, flexible and portable.
- LightGBM is a gradient boosting framework that uses tree based learning algorithms.
- Common parameters:
  - objective specify the learning task and the corresponding learning objective;
  - eval\_metric evaluation metrics for validation data
  - learning\_rate The parameter controls the magnitude of this change in the estimates.
  - max\_depth This indicates how deep the built tree can be.

## XGBoost parameters

- n\_estimators represents the number of trees in the forest
- min\_samples\_split represents the minimum number of samples required to split an internal node
- min\_samples\_leaf min\_samples\_leaf is The minimum number of samples required to be at a leaf node
- max\_features represents the number of features to consider when looking for the best split
- Subsample subsample ratio of the training instance
- colsample\_bytree subsample ratio of columns when constructing each tree.
- min\_child\_weight minimum sum of instance weight(hessian) needed in a child.
- Gamma minimum loss reduction required to make a further partition on a leaf node of the tree.

Source: https://xgboost-clone.readthedocs.io/en/latest/parameter.html

## LightGBM parameters

**sub\_feature -** will randomly select part of features on each iteration if feature\_fraction smaller than 1.0

num\_leaves - max number of leaves in one tree

min\_data - minimal number of data in one leaf.

max\_bin - max number of bins that feature values will be bucketed in

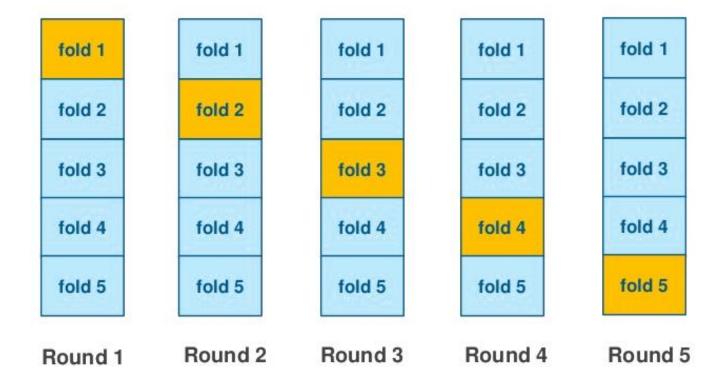
bagging\_freq - frequency for bagging

Source: https://lightgbm.readthedocs.io/en/latest/Parameters.html

## Choosing the parameters

- Understanding the data
- Understanding the parameter influence to the output
- Validating the model
- Hyper parameter optimization E.g. Sklearn Gridsearchcv
- Cross validation e.g. Sklearn cross\_validate

## **Cross validation**



score(CV) = the average of evaluation scores from each fold You can also repeat the process many times!



## **Model validation**

#### Training Set

this data set is used to adjust the weights

#### Validation Set

this data set is used to minimize overfitting.

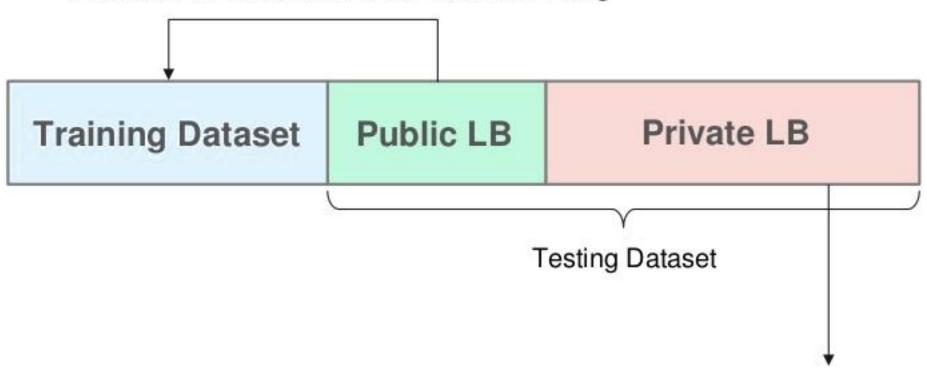
#### Testing Set

 this data set is used only for testing the final solution in order to confirm the actual predictive power of the network.

Important! Kaggle: public vs private leader boards

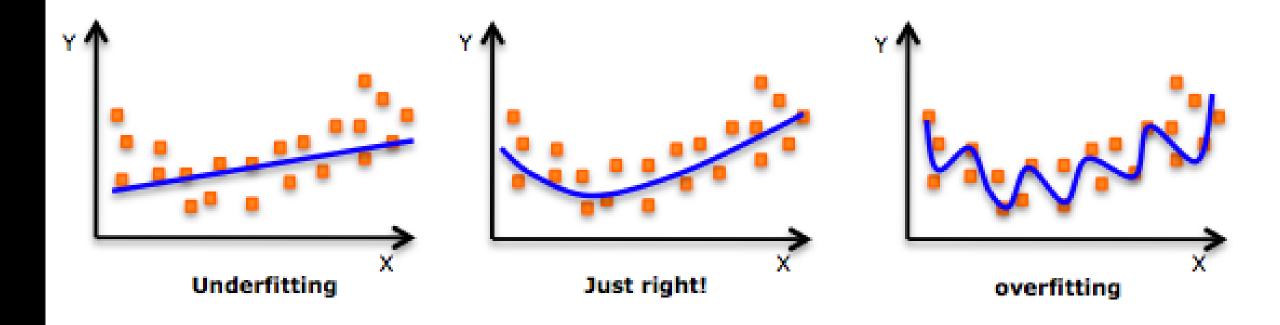
## Why validation is important?

Validation feedback but sometimes misleading



Might be different from public LB (used to determine final prize winners!)

## Model must be able to generlize



## Feature importance

- Move back to step 1: Data exploration
- Use model feature importance libraries e.g. sklearn feature\_selection
  SelectFromModel, Shap interpreter
- Feature importance determination uses different approaches, e.g. R2

#### Additional resources

- Basic courses for excel and so on, <a href="https://www.lynda.com/">https://www.lynda.com/</a>
- Deep learning course , <a href="https://www.fast.ai/">https://www.fast.ai/</a>
- Dive into Deep Learning, <a href="https://d2l.ai/">https://d2l.ai/</a>
- Agent-based modelling, complexity approach, <a href="https://www.complexityexplorer.org/">https://www.complexityexplorer.org/</a>
- Micromasters programmes, <a href="https://www.edx.org/">https://www.edx.org/</a>
  - Program in Supply Chain Management, Massachusetts Institute of Technology
  - Business and Economics for a Circular Economy, Wageningen University & Research
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   A highly efficient gradient boosting decision tree. In *Advances in Neural Information Processing Systems* (pp. 3146-3154).
- Zhou, Z. H., & Feng, J. (2017). Deep forest: Towards an alternative to deep neural networks. *arXiv preprint arXiv:1702.08835*.

# Working

- 1. Estimate the given real estate price:
  - Use file in data folder New customer request.xlsx
- 2. Improve the RMSE metric and submit to kaggle:
  - use LightGBM notebook (LightGBM implementation)
  - Implement XGBoost (Implement XGboost, Step by step)