# **Machine Learning Engineer Nanodegree**

## **Capstone Project**

# P6: Sberbank Russian Housing Market

## **Capstone Proposal**

### **Domain Background**

**Regression analysis** is a form of math predictive modeling which investigates the relationship between variables. It answers the questions: Which factors matter most? Which can we ignore? How do those factors interact with each other? And, perhaps most importantly, how certain are we about these factors and their predicitons?

The main factor that we're trying to understand or predict is a target (a dependent variable). The features (independent variables) are the factors we suppose to have an impact on the dependent variable. Using this set of variables, we generate a function that map inputs to outputs. The training process continues until the model achieves the desired level of accuracy.

The project investigates **supervised learning** as a part of regression analysis that uses a known dataset (the training dataset) to make predictions. The training dataset includes input data and response values. The supervised learning algorithms seek to build models which make predictions of the response values for a new dataset. A test dataset is used to validate the model.

**Housing costs** are a sphere in the real economy for applying supervised learning. They demand a significant investment from both consumers and developers. And when it comes to planning a budget—whether personal or corporate—the last thing anyone needs is uncertainty about one of their budgets expenses. Sberbank, Russia's oldest and largest bank, helps their customers by making predictions about reality prices so renters, developers, and lenders are more confident when they sign a lease or purchase a building.

Although the housing market is relatively stable in Russia, the country's volatile economy makes forecasting prices as a function of apartment characteristics a unique challenge. Complex interactions between housing features such as a number of bedrooms and location are enough to make pricing predictions complicated. Adding an unstable economy to the mix means Sberbank and their customers need more than simple regression models in their arsenal.

#### **Problem Statement**

Sberbank is challenging programmers to develop algorithms which use a broad spectrum of features to predict real prices. Algorithm applications rely on a rich dataset that includes housing data and macroeconomic patterns. An accurate forecasting model will allow Sberbank to provide more certainty to their customers in an uncertain economy.

### **Datasets and Inputs**

The basis for the investigation is a large number of economic indicators for pricing and prices themselves (train.csv and test.csv). Macroeconomic variables are collected in a separate file for

transaction dates (macro.csv). In addition, the detailed description of variables is provided (data\_dictionary.txt).

For practical reasons, I have not analyzed all the data and have chosen the following independent variables:

- 1. the dollar rate, which traditionally affects the Russian real estate market;
- 2. the distance in km from the Kremlin (the closer to the center of the city, the more expensive);
- 3. indicators characterizing the availability of urban infrastructure nearby (schools, medical and sports centers, supermarkets, etc.);
- 4. indicators of a particular living space (number of rooms, floor, etc.);
- 5. proximity to transport nodes (for example, to the metro);
- 6. indicators of population density and employment in the region of housing accommodation.

#### **Solution Statement**

The project solutions consist of two main parts:

- 1. preparation of data for analysis (selection of variables, deletion of records containing too many empty values, digital encoding categorical variables, etc.);
- 2. application of a set of machine learning algorithms in regression analysis in order to identify the most effective of them.

#### **Benchmark Models**

- 1) Scikit Learn: Gradient Boosting Regressor, Bagging Regressor, MLP Regressor;
- 2) Keras: multi-layer perceptrons (MLP), convolutional neural networks (CNN), recurrent neural networks (RNN).

#### **Evaluation Metrics**

The wide spectrum of metrics for regression was chosen and documented:

- 1. explained variance regression score;
- 2. coefficient of determination;
- 3. mean squared error;
- 4. mean absolute error;
- 5. median absolute error.

## **Project Design**

The project was built on the basis of the competition offered on the site https://www.kaggle.com.

Here popular Python resources (numpy, pandas, matplotlib, scikit-learn, keras, etc.) for building the regression models were applied.

The most valuable side of this project is the investigation of real data and the attempt to approximate the predictions on them to the threshold of 70-80 percentages.

# Bibliography

- 1. Amy Gallo. A Refresher on Regression Analysis. Harvard Business Review, 2015.
- 2. Model evaluation: quantifying the quality of predictions (<a href="http://scikit-learn.org/stable/modules/model">http://scikit-learn.org/stable/modules/model</a> evaluation.html)
- 3. Keras: The Python Deep Learning library (<a href="https://keras.io/">https://keras.io/</a>).