

# Machine Learning Engineer Nanodegree

## Capstone Proposal

Anton Mironenko  
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### Proposal

House prices prediction, from Kaggle competition

### Domain Background

Quite often it is a challenge to predict house prices based on a big number of factors. Such a model with a decent accuracy would be a good tool for real estate agents. Machine Learning provides a lot of tools to solve the tasks like this.

There is a good corresponding competition on Kaggle with a clean dataset:  
<https://www.kaggle.com/c/house-prices-advanced-regression-techniques>.

My personal motivation is to apply knowledge and techniques gained from Udacity Machine Learning Nano Degree to some practical project.

### Problem Statement

Predicting house prices is a regression problem, where the output variable takes continuous values.

### Datasets and Inputs

The following public dataset is going to be used:

<https://www.kaggle.com/c/5407/download-all>

It has 80 features, 1 dependent output variable (price), 1460 training data points and 1459 test data points.

### Solution Statement

The solution will use regression methods with a help of Python and required machine learning libraries. It shall predict the price based on input variables.

## Benchmark Model

The dataset is taken from Kaggle ongoing competition, so that its results can be taken as a benchmark model.

The goal is to provide a model with a metric fit in top 50% of Kaggle competition's results.

## Evaluation Metrics

The Kaggle competition uses Root-Mean-Squared-Error (RMSE) between the logarithm of the predicted value and the logarithm of the observed sales price: <https://www.kaggle.com/c/house-prices-advanced-regression-techniques#evaluation>

The same metric will be used in the solution model.

## Project Design

The following steps are going to be made to solve the Problem:

- Visualize and analyze the dataset, find correlations and other dependencies,
- Clean and normalize the dataset:
  - remove outliers,
  - fill missing data,
  - transform numeric features if needed,
  - perform one-hot-encoding for categorical features,
  - reduce dimensionality: filter out less important features.
- Try different regression methods, for example: LinearRegression, Lasso, Ensemble methods (AdaBoost, GradientBoost, RandomForest). Probably, some popular on Kaggle methods to be evaluated: XGBoost, LightGBM.
- Analyze obtained evaluation metrics, perform retrospective analysis.