**Cars Prices Forecasting**

**Motivation**

Understanding of the principles of cars prices formation is important for customers, dealers and intermediaries. If you are buying a brand new car from an authorized dealer, you face with sticky prices. However, if you want to buy used car from an individual or unauthorized dealer, prices can vary sharply. If you want to sell your car, you probably do not know the exact price of your vehicle on aftermarket. Car price obviously depends on broad variety of different parameters and it is very hard to keep all in mind if you are not a car reseller and selling cars is not your business. Ability to forecast car price based on its characteristics could save a lot of money for customers and increase profits of dealers and intermediaries.

I also have personal motivation in this project because in the nearest future I plan to buy a car.

**Data collection:**

To start I limited variety of cars for analysis to most popular D-segment cars in Russia[[1]](#footnote-1). <https://moscow.auto.ru/> – one of the most visited automotive sites in the Russian internet was selected as data source for analysis. This website has most of cars selling advertisements of new and used cars. All materials for this project can be found at <https://github.com/BiXiC/auto_ru>.

1. To scrap webpages with cars advertisements, I used Python 3 and BeautifulSoup library and wrote the script [**AutoRu\_parser.ipynb**](http://localhost:8889/notebooks/AutoRu_parser.ipynb) that parse all advertisements from Moscow and +200 km area around it and saves data to pickle files (auto\_ru/data/D\_class).
2. For preparation of features matrix, I wrote **features\_calculation.ipynb** that cleans data and extracts some useful features from scrapped data.

1. Short list of models for analysis: Kia Optima, Honda Accord, Hyundai i40, Hyundai Sonata, Toyota Camry, Mazda 6, BMW 5er, Audi A6, Ford Mondeo, Infiniti G35, Nissan Teana, Opel Insignia, Volvo S60 [↑](#footnote-ref-1)