## YYDS Group Project

House Prices of Beijing, Shanghai and Guangzhou

# Group Member Introduction



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## Reason topic was selected



After graduation, we may face the problem of buying a house. So, we are curious about the house price in China, and also the trend for it.

With our analysis, we know what kind of house is cheaper and we can predict the house price of any kind of houses.

## Description of the source of data



We selected some data of Beijing House Price from Kaggle

https://www.kaggle.com/ruiqurm/lianjia

All Data in Kaggle is from the following website

https://bj.lianjia.com/chengjiao.

but thats not enough, to make our analysis more complex and meaningful, we did a lot of web scraping and got House Price in Shanghai and Guangzhou as well.

Moreover, we not only considered houses that already been sold, but we also considered houses that are currently on sale.

# Questions the team hopes to answer with the data



- 1. Which city's houses are more expensive.
- 2. Which district's houses are more expensive.
- 3. Create interactive tables and charts for customers and our audience to select their dreaming house.
- 4. Predict house price for customer's dreaming house.

# Description of the data exploration phase of the project



- Discussing where to find the big data, and which data to choose
- 2. Checking the complexity of the data
- 3. Seeking more data
- 4. Data cleaning
- 5. Create the database

# Technique used in this project



#### 1.Python

- Web Scrapping (pandas, numpy, BeautifulSoup, ChromeDriverManager, etc)
- Data Cleaning (sqlalchemy, config, request, etc)
- Machine Learning (sklearn, matplotlib, seaborn, sqlite3, etc)
- Flask

#### 2.JavaScript

- HTML
- Dynamic Tables
- Data Visualization (Charts)
- Mapping with JS and APIs

#### 3.SQL

- SQLite

## Machine Learning



- Supervised Learning: Regression
- Outcome(y): Total Price
- Features(X):
  - City
  - District
  - FloorLevel
  - FloorType
  - BuildingType
  - BuildingStructure
  - RenovationCondition
  - Elevator
  - Number of Bedrooms
  - Number of Living rooms
  - Number of Kitchen
  - Number of Bathrooms
- LabelEncoder, StandardScaler
- Split 80% data into training set and 20% data into testing set

### Machine Learning



LinearRegression Mean Squared Error: 85869.92055431839

LinearRegression R2: 0.38539979917493605

DecisionTreeRegressor Mean Squared Error: 55600.36686514736

DecisionTreeRegressor R2: 0.6020492808113085

RandomForestRegressor Mean Squared Error: 40721.46958394869

RandomForestRegressor R2: 0.7085426046440196

LGBMRegressor Mean Squared Error: 36747.29619343738

LGBMRegressor R2: 0.7369871140619224

SVR Mean Squared Error: 149021.97425272633

SVR R2: -0.06660090881459557

KNeighborsRegressor Mean Squared Error: 55938.968084646935

KNeighborsRegressor R2: 0.5996257968234271
Lasso Mean Squared Error: 85869.90741591265

Lasso R2: 0.38539989321097046

The result shows LightGBM as our best model since it has the lowest MSE and the highest R2.

### DashBoard



Now let's have a look at our beautiful, gorgeous, remarkable, splendid, amazing dashboard :)))

## We aim to create a dashbaord that can link to different pages.

- We will have home page, about us, team, page information for different cities, mapping, etc.
- We use javascript to make the dashboard
- We plan to add what we learn from the module (searching bar, mapping, graphing)

# Thank You

Feel free to ask any questions:)