

STRUCTURE

- Business values and metrics
- Modeling
- Pipeline
- API
- Q&A



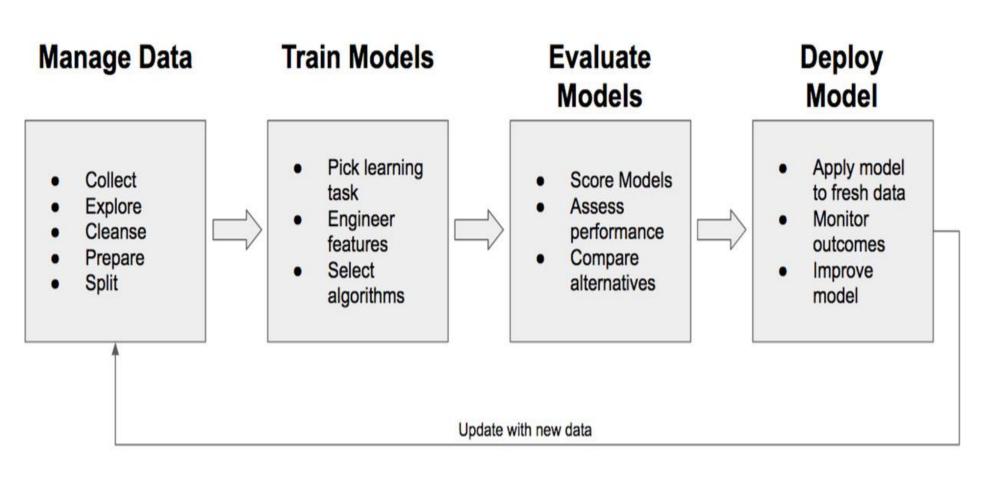


INTRODUCTION

Accurate housing price prediction help investors identify potential investment opportunities in real estate business

The aim of the project is to build a machine learning model pipeline to predict the sale price of homes

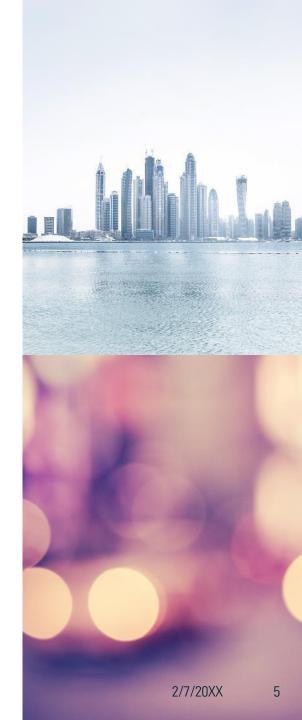
Machine Learning Modeling Cycle





MACHINE LEARNING PIPELINE

- 1. Data Analysis
- 2. Feature Engineering
- 3. Feature Selection
- 4. Model Training
- 5. Predictions Scoring
- 6. API
- 7. CI/CD
- 8. DOCKER



DATA ANALYSIS

Train.csv / Test.csv

The house price dataset contains 2051 rows, 80 columns:

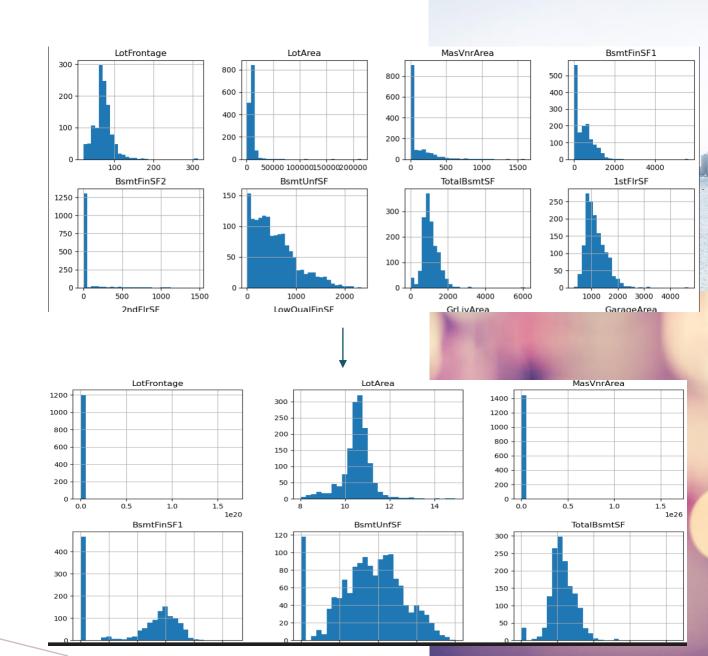
- 79 predictive variables
- l target: SalePrice



DATA ANALYSIS

DATA TRANSFORMATION

- Yeo-Johnson
- Log
- Skewed >> binary



FEATURE ENGINEERING

- 1. Missing values
- 2. Temporal variables
- 3. Non-Gaussian distributed variables
- 4. Categorical variables: remove rare labels
- 5. Categorical variables: convert strings to numbers
- 5. Put the variables in a similar scale
- 6. Mapping assigned order
- 7. Feature scaling

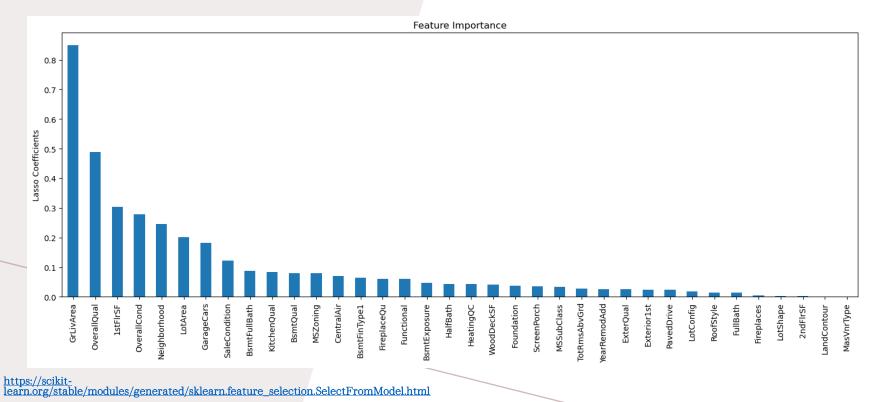


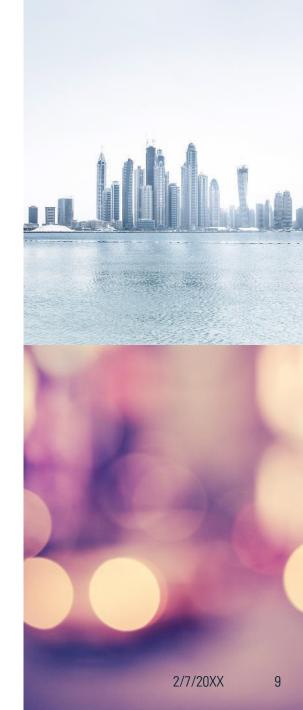
FEATURE **SELECTION**

Use scikit learn: selectfrommodel

total features: 80

selected features: 36



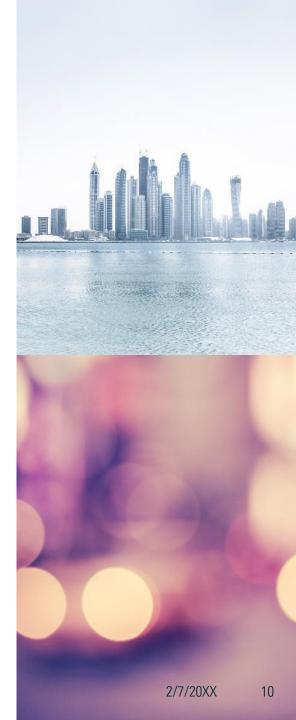


MODEL TRAIN

RMSE: 18179.7065

Model comparisons

- Linear regression
- Lasso
- Ridge
- ElecticNet
- Random Forest
- XGBoost



SCORING

	TRAIN	TEST
RMSE	27769	33016
R2	0.876496042007156	0.8413724432687755



LINEAR REGRESSION

Model_pipeline.py

- Feature pipeline
- Linear_model
- Score test data

joblib.dump(lin_model,
 'linear_regression.joblib')



FAST API

Easy to implement

Product codes being tested with TOX https://tox.wiki/en/latest/

http://localhost:8001/

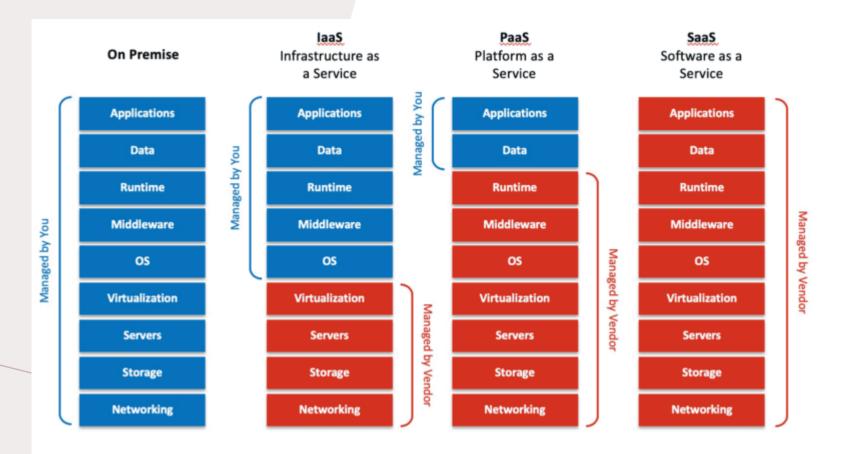
Demo

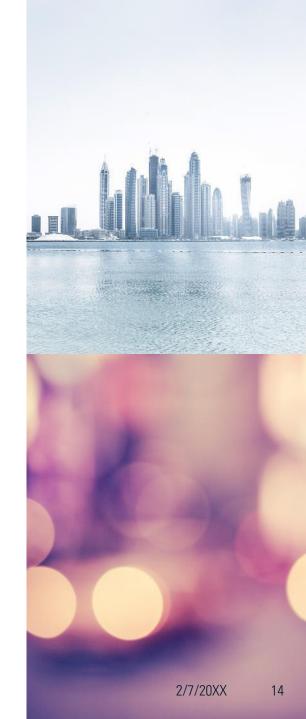
```
■ 04-machine-learning-pipeline-model-training.ipynb M ● ≡ tox.ini M X
       [testenv:run_app]
       envdir = {toxworkdir}/test_app
          {[testenv:test app]deps}
          python app/main.py
PS C:\Users\chend\Documents\mlops\deploying-machine-learning-models\section-05-production-model-package> tox -e test_package test_package: commands[0]> python regression model/train_pipeline.py test_package: commands[1]> pytest -s -vv tests/
cachedir: .tox\test_package\.pytest_cache
rootdir: C:\Users\chend\Documents\mlops\deploying-machine-learning-models\section-05-production-model-package, configfile: pyp
tests/test_features.py::test_temporal_variable_transformer PASSED
tests/test_prediction.py::test_make_prediction_PASSED
```



CLOUD ENVIRONMENT

Platform as a service (PaaS)





RAILWAY

Railway is an infrastructure platform where you can provision infrastructure, develop with that infrastructure locally, and then deploy to the cloud

500 hours free

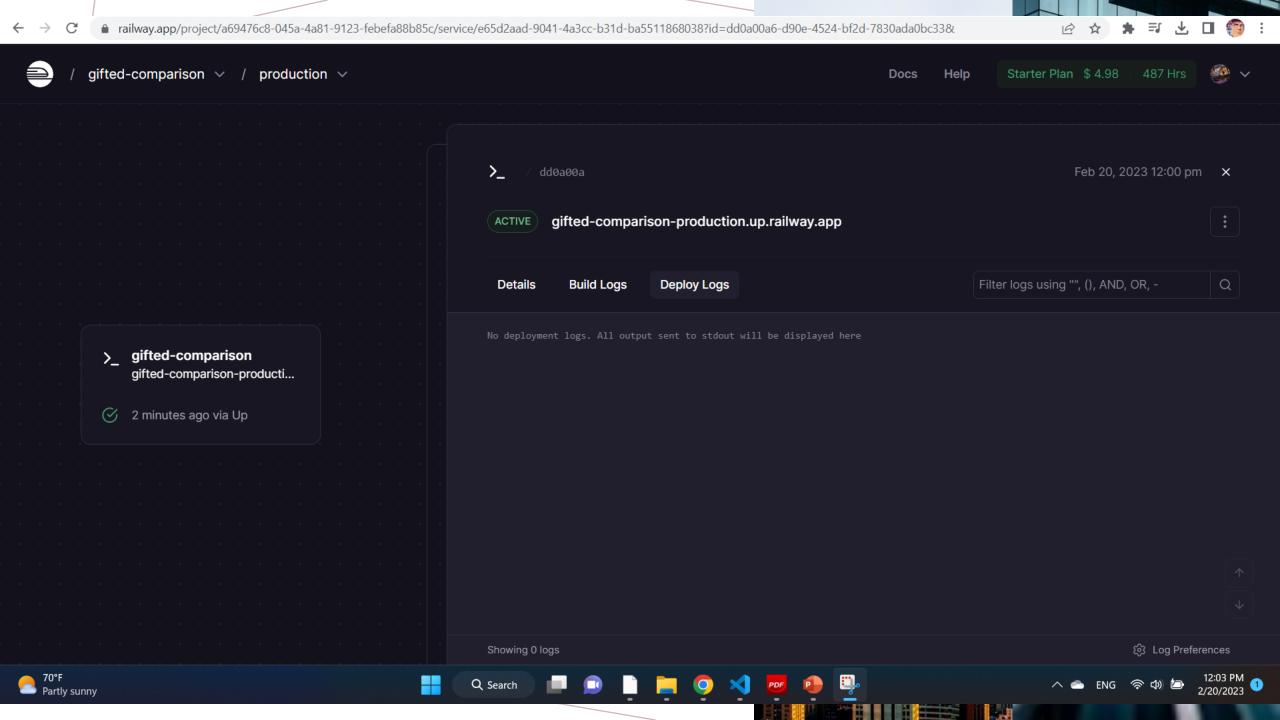
Support large dependency size on free tier

Work with docker

Github integration





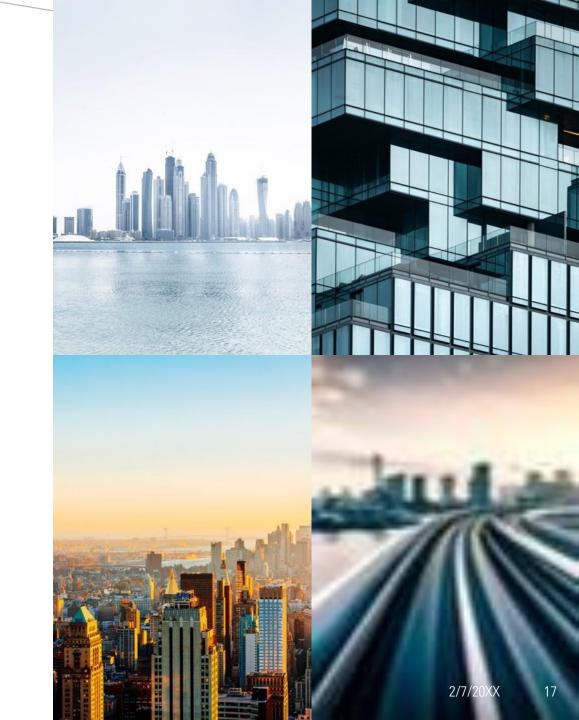


CI/CD

CircleCI is a continuous integration and continuous delivery platform that can be used to implement DevOps practices

Alternatives

- Jenkins
- Gitlab ci
- Travis CI

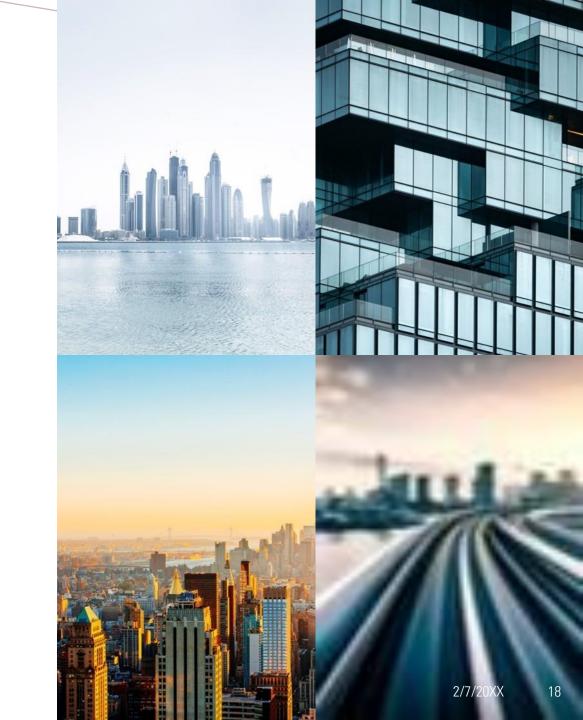


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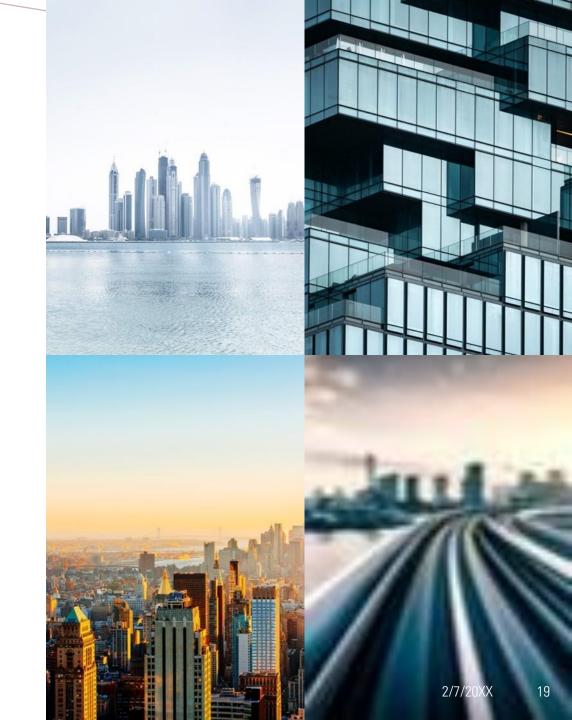
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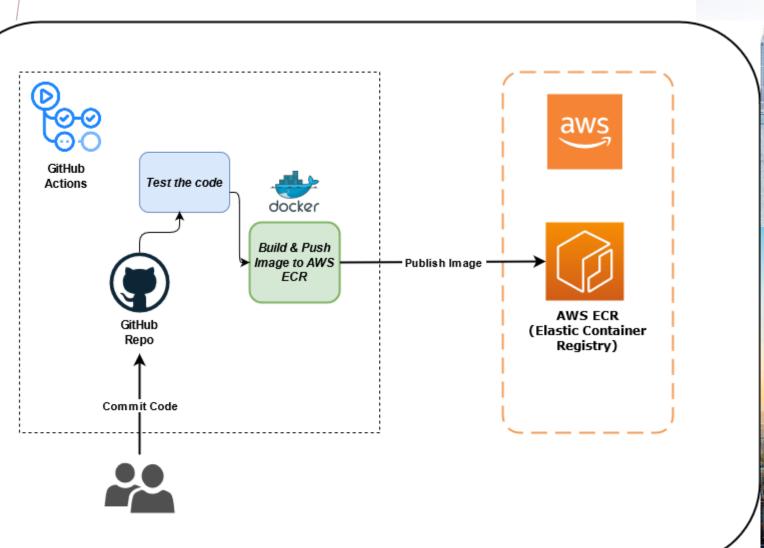
IAAS

Infrastructure as a Service (IaaS) is a business model that delivers IT infrastructure like compute, storage, and network resources on a pay-as-you-go basis over the internet



AWS ENVIRONMENT

Calliple Public Text





OUTCOME

- 1. Built a prediction model with low RMSE that can predict price with new data
- 2. Deploy the model serve from FAST API
- Deploy model in production environment RAILWAY
- 4. CI/CD to cloud platform with Docker

