

# Machine Learning Canvas

## PREDICTIONS

### End-user

Who will use the predictive system / who will be affected by it?

- Primary users: Equity Residential property management team, real estate market analysts, marketing and sales department.
- Secondary users: Potential tenants, ...

## OBJECTIVES

### Value proposition

What are we trying to do for the system's users? (e.g. spend less time on X, increase Y...)

- Increase the accuracy of rent price forecasting.
- Maximize rental income.
- Decrease vacancy rates.
- Enhance tenant satisfaction through competitive and transparent pricing

## DATA

### Data sources

Where do/can we get data from? (internal database, 3rd party API, etc.)

Web scraping data from the Equity Residential website and other rental property websites.

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## PREDICTIONS

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### Problem

Question to predict answers to (on behalf of user)

Question: What will be the optimal rent for a specific ap

Input (i.e. question "parameter")

- Price - Beds - Baths - sq.ft - Floor - Move\_in\_date - bu

Possible outputs (i.e. "answers")

Predicted rent price.

Type of problem (e.g. classification, regression, recommendation...)

Regression.

Baseline: simple, alternative way of making predictions (e.g. manual rules)

Manual rules based on historical data and current market conditions.

### Performance evaluation

Domain-specific / bottom-line metrics for monitoring performance in production

- Vacancy rate.
- Rental income.

Prediction accuracy metrics (e.g. MSE if regression; % accuracy, #FP for classification)

- Mean Absolute Error (MAE)
- Mean Squared Error (MSE)
- Root Mean Squared Error (RMSE).
- Coefficient of Determination ( $R^2$ ).

Offline performance evaluation method (e.g. cross-validation or simple training/test split)

- Cross-validation.
- Simple train-test split.

### Data preparation

How do we get training data (inputs, and outputs if supervised learning)? How many data points?

- Historical data obtained via web scraping from Equity Residential.
- Amount of data: 62,810 records.

Input features (extracted from data sources). If too many, list types of features and mention key ones.

- Balcony
- Walk\_In\_Closet
- Fireplace
- City\_Skyline
- Kitchen\_Island
- Stainless\_Appliances
- Renovated
- Office\_Space
- Days\_Till\_Available
- Day\_of\_the\_week\_recorded
- Unique\_ID

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### Using predictions

When do we make predictions and how many?

- Predictions are made upon receiving a new listing and during weekly data updates.
- The number of predictions varies depending on the number of listings.

What is the time constraint for making those predictions?

Predictions should be made in real-time or within a few seconds after the request.

### Learning models

When do we create/update models? With which data / how much?

- Models are updated annually or when there are significant market changes.
- Data from the past year is used for model updates.

What is the time constraint for creating a model?

The model should be created within a few hours to ensure timely updates.

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## PREDICTIONS

How do we use predictions and confidence values?

- Predicted rent prices are used for setting rates on the website.
- Confidence values can be used for automatic rate checking and adjustment, as well as for management decision-making.

## OBJECTIVES

Criteria for deploying model (e.g. minimum performance value — absolute, relative to baseline or to previous model)

- Minimum model performance:  $MAE < \$50$ .
- $R^2 > 0.85$ .
- Performance must be better than or equal to the baseline model and previous versions.

## DATA

Reset Form

Machine Learning Canvas v0.1

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