

# Executive summary



# **Motivation & goals**



## **Deliverables**

Data set
Analytics Dashboard
Deal Fairness Indicator



# **Next steps**

Challenges & limitations
Future improvements

# Utilizing analytics tools and machine learning reduces search time by up to 50%

### Goal



## Approach

Develop analytical dashboard



Narrow down option

Price Fairness Meter

## Example



Great deal! This apartment is listed at 10% lower than its fair value.

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# We gathered data of the majority of apartments currently listed for sale in Riyadh, amounting to 7000+ apartments

#### **Attributes** • # of Living/Guest rooms # of Bedrooms **Floor Plan** # of Bathrooms Has Kintchen · Length:Width Ratio • Area (sqr. m) Area Shape Regularity Dimensions Street Direction District Location Latitude & Longitude Street Width Installed AC Installed Kitchen **Features** Furnished Age **Building** Apartment Floor Advertisor Type (owner) Time on Market Post ID **Online Post** Advertisor Rating Last Update Advertisor ID

#### **Conclusions**

- 23+ gathered and engineered attributes for each apartment
- Some attributes suffer greater levels of inaccuracy
- Data was gathered by scraping Aqar.com

Agar only offers filtering search results by price and location, which results in thousands of results; we enable our users to target their efforts by filtering for exactly what they want



# We developed a Deal Fairness Meter to help users determine if an apartment they are interested in has a reasonable asking price

Get assessment for an apartment's price! Enter the apartment post ID:



Submit

This is a FAIR DEAL. Actual asking price is 800k. Our estimate of a fair price for this apartment is: 788k

#### **Approach to generate Deal Fairness Score**



- Search Agar.com to find the apartment with the given Post ID
- Collect the apartment data and prepare it to enter the machine learning model



- Use the data as input into the ML model
- Output is our estimate of a reasonable price



- Compare our generated reasonable price with actual asking price
- Report: {Terrible Deal, Bad Deal, Fair Deal, Good Deal, Great Deal}



- Additional step: run all apartments in the database through the Deal Fairness Meter
- · Allow users to filter for only Good and Great Deals

#### Notes

- Our estimate for a "reasonable price" is based on the 23+ attributes of each apartment
- 7000+ apartments were used to train the machine learning model
- The model's predictions are accurate, with an average prediction error of ~20k SAR
- The tool is currently limited to searching for listings on Agar.com

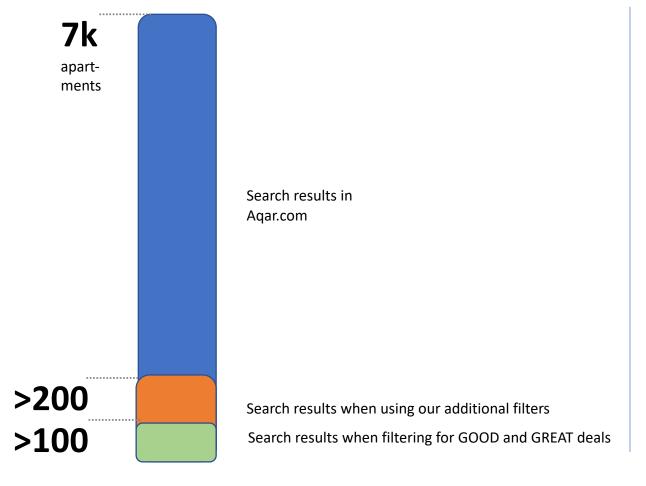
#### The Deal Fairness Meter is used for two tasks:

- Enable users to access outside advice for an apartment they found online which they are interested in
- Enable users to quickly filter our database for only the good deals, then download that list to find those listings on Aqar.com

After narrowing down the search results to a managable number, the user can download a list of candidate apartments for further analysis

| A                 | В       | С             | D    | Е       | F  | G    | н            | ı   | J             | K       | L         |
|-------------------|---------|---------------|------|---------|----|------|--------------|-----|---------------|---------|-----------|
| 1 post_id         | price   | link          | beds | livings | wc | area | street_width | age | street_direct | ketchen | furnished |
| 2 3963975         | 850000  | https://sa.aq | 3    | 1       | 3  | 129  | 25           | 16  | - 5           | 1       | 0         |
| 3 4430695         | 1400000 | https://sa.aq | 3    | 2       | 3  | 142  | 30           | 1   | 2             | 1       | 0         |
| 4 4420689         | 1200000 | https://sa.aq | 3    | 2       | 3  | 166  | 30           | 1   | 2             | 1       | 0         |
| 5 3483081         | 575000  | https://sa.aq | 3    | 3       | 4  | 170  | 30           | 0   | 3             | 0       | 0         |
| 6 4391509         | 890000  | https://sa.aq | 2    | 1       | 2  | 124  | 36           | 0   | 3             | 1       | 0         |
| 7 3595328         | 1150000 | https://sa.aq | 3    | 1       | 3  | 164  | 30           | 0   | 1             | 0       | 1         |
| 8 4438114         | 1300000 | https://sa.aq | 4    | 1       | 3  | 220  | 36           | 0   | 1             | 1       | 0         |
| 9 4190651         | 970000  | https://sa.aq | 4    | 1       | 3  | 182  | 20           | 0   | 3             | 0       | 0         |
| 10 4443656        | 1199000 | https://sa.aq | 3    | 1       | 4  | 145  | 30           | 0   | 5             | 0       | 0         |
| 11 4190548        | 920000  | https://sa.aq | 4    | 1       | 3  | 200  | 30           | 0   | 4             | 0       | 0         |
| 12 4359292        | 830000  | https://sa.aq | 4    | 1       | 3  | 137  | 30           | 0   | 3             | 0       | 0         |
| 13 3900139        | 1000000 | https://sa.aq | 5    | 1       | 2  | 132  | 18           | 9   | 6             | 1       | 1         |
| 14 4300375        | 1170000 | https://sa.aq | 3    | 2       | 3  | 176  | 36           | 0   | 8             | 1       | 0         |
| 15 4075233        | 1300000 | https://sa.aq | 3    | 1       | 3  | 129  | 36           | 0   | 1             | 0       | 0         |
| <b>16</b> 4583365 | 950000  | https://sa.aq | 4    | 1       | 3  | 199  |              | 1   | 9             | 0       | 0         |
| 17 4369844        | 1149000 | https://sa.aq | 4    | 1       | 3  | 163  | 30           | 0   | 2             | 1       | 0         |
| 18 3766687        | 980000  | https://sa.aq | 3    | 2       | 3  | 135  | 36           | 0   | 4             | 0       | 0         |
| 19 4390778        | 1600000 | https://sa.aq | 3    | 1       | 3  | 130  | 32           | 0   | 4             | 1       | 0         |
| 20 4321372        | 1050000 | https://sa.aq | 3    | 1       | 3  | 172  | 30           | 0   | 4             | 1       | 0         |
| 21 3524161        | 1200000 | https://sa.aq | 3    | 1       | 3  | 148  | 36           | 0   | 4             | 1       | 0         |
| 22 4450332        | 929000  | https://sa.aq | 3    | 1       | 3  | 154  | 30           | 0   | 4             | 0       | 0         |
| 23 4127891        | 1200000 | https://sa.aq |      | 1       | 3  | 146  | 36           | 0   | 4             | 1       | 0         |
| 24 4352329        |         |               |      | 2       | 4  | 160  | 25           | 4   | 4             | 1       | 1         |
| 25 3331722        | 1950000 | https://sa.aq | 2    | 2       | 2  | 90   | 15           |     | 5             | 1       | 1         |

With our filteration system and Deal Fairness Scores, we can narrow down search results by more than 90%, saving users 50+ hours of search



#### Value created

- On average, home-buyers spend 100 hours of search
- If our services saves 50% of that search time, then we are saving our users 50 hours of work
- Assuming an hour is worth 5 SAR (certainly an under-estimate), then we are creating 250 SAR of value for each customer!

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# We recognize the huge value generated by this tool, and we recognize its many aspects that need further development

#### Limitations



 Our database and Deal Fairness Meter tool are limited to apartments listed in Agar.com

#### **Future developments**

 Develop data-scraping scripts for other platforms



Some listings contain false information about their properities, which hinders our ability to generate accurate predictions

 Develop machine learning models that can perform well desbite missing data



Our prediction model performs particularly poorly on some types of apartments (e.g., high-end luxury apartments)

 Develop multiple machine learning models for the different price ranges



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