# Module 2 Data Science Project King County property price predictor

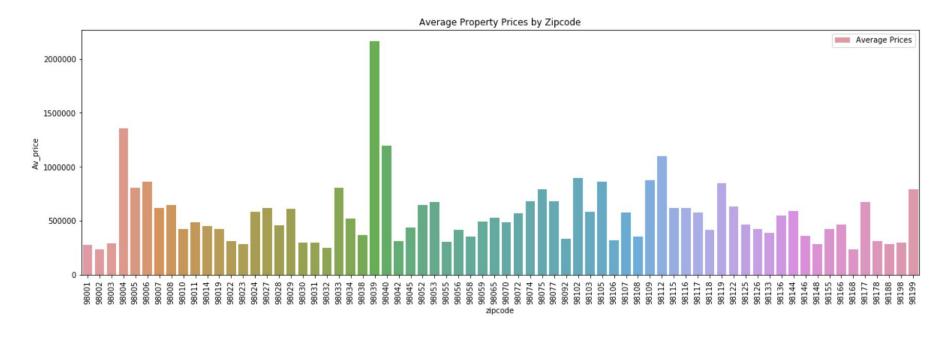
By Naweed and Jim

#### **Intro & Project Outcomes**

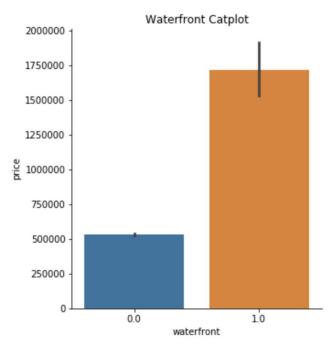
Our aim with this project is to produce effective pricing models for the domestic housing market within the King County area. We have investigated a number of parameters available to us to give accurate predictions of house prices in different cases. We have been guided by the demand for information by different sections of the domestic housing market - homeowners, developers, investors and agents looking for more insight. We developed the following guide to inform our work:

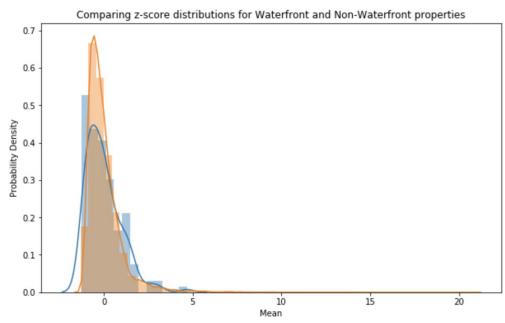
- 1. Create and demonstrate an accurate tool to predict house prices within the King County Area by zip-code.
- 2. Provide a guide to current homeowners who are considering adding value to their property and would like to understand if it is worth the initial outlay.
- 3. Provide a guide to prospective homeowners who would like to know whether a property is under or over-valued.

#### How do average property prices rank across King County?

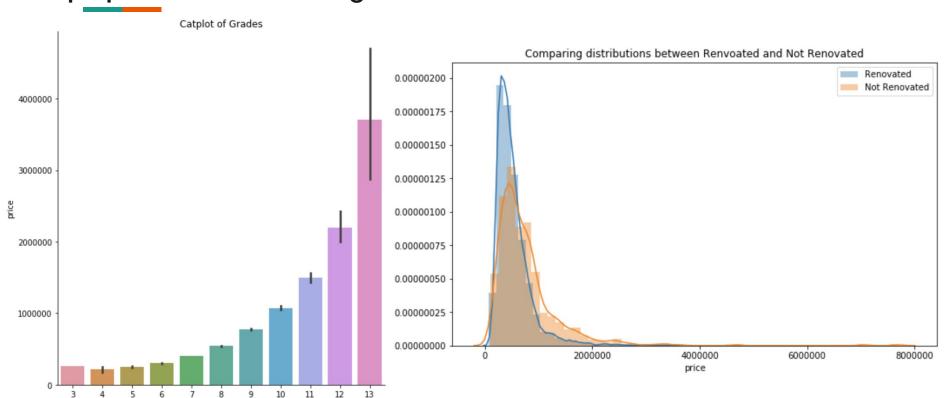


### Does living by the waterfront add value?





## Does Grade add value to property? If so is it worth buying low grade properties and renovating?



#### Final Regression Model

$$price = 293 * \beta_{sqft_living} + 124,077 * \beta_{renovate_15} + 26,368 * \beta_{waterfront} - 26,368 * \beta_{yr_built} + 4,100,000$$

| OLS Regression Results |                  |             |              |           |                |           |
|------------------------|------------------|-------------|--------------|-----------|----------------|-----------|
| Dep. Vari              | able:            | price       | R-squared:   |           | <b>d:</b> 0    | .559      |
| Me                     | odel:            | OLS         | Adj. R       | -square   | <b>d:</b> 0    | .559      |
| Met                    | t <b>hod:</b> Le | ast Squares | F-statistic: |           | <b>c:</b> 6    | 853.      |
| ı                      | Date: Fri, 2     | 27 Mar 2020 | Prob (F-     | statistic | ):             | 0.00      |
| Т                      | Time:            | 03:33:12    | Log-Li       | kelihoo   | d: -2.9851e    | e+05      |
| No. Observat           | ions:            | 21595       |              | Ale       | 5.970e         | e+05      |
| Df Resid               | uals:            | 21590       |              | ВІ        | <b>5</b> .971e | e+05      |
| Df M                   | odel:            | 4           |              |           |                |           |
| Covariance T           | уре:             | nonrobust   |              |           |                |           |
|                        | coef             | std err     | t            | P> t      | [0.025         | 0.975]    |
| const                  | 4.103e+06        | 1.18e+05    | 34.828       | 0.000     | 3.87e+06       | 4.33e+06  |
| sqft_living            | 293.9264         | 1.929       | 152.348      | 0.000     | 290.145        | 297.708   |
| renovate_15            | 1.241e+05        | 1.28e+04    | 9.692        | 0.000     | 9.9e+04        | 1.49e+05  |
| waterfront             | 8.188e+05        | 2.04e+04    | 40.132       | 0.000     | 7.79e+05       | 8.59e+05  |
| yr_built               | -2121.9655       | 60.396      | -35.134      | 0.000     | -2240.347      | -2003.584 |

#### **Baseline Regression Model**

$$price = 270 * \beta_{sqft_living} + 862,256 * \beta_{waterfront} + 26,363 * \beta_{basement} - 37,390$$

|                  |               | OLS Regression Results |        |            |          |          |             |          |  |  |
|------------------|---------------|------------------------|--------|------------|----------|----------|-------------|----------|--|--|
|                  | Dep. Varia    | ble:                   |        | price      | R        | -square  | ed:         | 0.531    |  |  |
|                  | Мо            | del:                   |        | OLS        | Adj. R   | -square  | ed:         | 0.531    |  |  |
|                  | Met           | hod:                   | Lea    | st Squares | F        | -statist | tic: 8      | 8139.    |  |  |
|                  | D             | ate:                   | Fri, 2 | 7 Mar 2020 | Prob (F- | statisti | c):         | 0.00     |  |  |
|                  | Т             | ime:                   |        | 10:43:45   | Log-L    | ikelihoo | od: -2.9921 | e+05     |  |  |
|                  | No. Observati | ons:                   |        | 21596      |          | Α        | IC: 5.984   | e+05     |  |  |
|                  | Df Residu     | ıals:                  |        | 21592      |          | В        | IC: 5.985   | e+05     |  |  |
|                  | Df Mo         | del:                   |        | 3          |          |          |             |          |  |  |
| Covariance Type: |               | уре:                   |        | nonrobust  |          |          |             |          |  |  |
|                  |               |                        | coef   | std err    | t        | P> t     | [0.025      | 0.975]   |  |  |
|                  | const -       | 3.739                  | e+04   | 4291.221   | -8.713   | 0.000    | -4.58e+04   | -2.9e+04 |  |  |
|                  | sqft_living   | 270.                   | 0042   | 1.914      | 141.083  | 0.000    | 266.253     | 273.755  |  |  |
|                  | waterfront    | 8.623                  | e+05   | 2.1e+04    | 41.021   | 0.000    | 8.21e+05    | 9.03e+05 |  |  |
|                  | basement      | 2.636                  | e+04   | 3593.738   | 7.336    | 0.000    | 1.93e+04    | 3.34e+04 |  |  |

## Test - Train Split (80/20 split)

#### **Train Dataset**

#### **Test Dataset**

| OLS Regression Results |                  |                     |             |  |  |  |
|------------------------|------------------|---------------------|-------------|--|--|--|
| Dep. Variable:         | price            | R-squared:          | 0.568       |  |  |  |
| Model:                 | OLS              | Adj. R-squared:     | 0.568       |  |  |  |
| Method:                | Least Squares    | F-statistic:        | 5681.       |  |  |  |
| Date:                  | Fri, 27 Mar 2020 | Prob (F-statistic): | 0.00        |  |  |  |
| Time:                  | 03:42:47         | Log-Likelihood:     | -2.3895e+05 |  |  |  |
| No. Observations:      | 17276            | AIC:                | 4.779e+05   |  |  |  |
| Df Residuals:          | 17271            | BIC:                | 4.779e+05   |  |  |  |
| Df Model:              | 4                |                     |             |  |  |  |
| Covariance Type:       | nonrobust        |                     |             |  |  |  |

|                   | OLS Regression Results |                     |           |  |  |
|-------------------|------------------------|---------------------|-----------|--|--|
| Dep. Variable:    | price                  | R-squared:          | 0.524     |  |  |
| Model:            | OLS                    | Adj. R-squared:     | 0.524     |  |  |
| Method:           | Least Squares          | F-statistic:        | 1189.     |  |  |
| Date:             | Fri, 27 Mar 2020       | Prob (F-statistic): | 0.00      |  |  |
| Time:             | 03:45:05               | Log-Likelihood:     | -59523.   |  |  |
| No. Observations: | 4319                   | AIC:                | 1.191e+05 |  |  |
| Df Residuals:     | 4314                   | BIC:                | 1.191e+05 |  |  |
| Df Model:         | 4                      |                     |           |  |  |
| Covariance Type:  | nonrobust              |                     |           |  |  |

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#### Recommendations

1. For anyone looking to invest in property in King County, it seems that waterfront properties do command a premium.

2. There seems to be a case for buying lower grade properties, renovating it and then selling it on.

#### **Future Work**

In order to expand our work going forward, it would be good to get access to data from a longer period to assess trends and make stronger recommendations.

It would be good to be able to get access to more location data and incorporate it into the model. As slide 3 shows, there does seem to be a few locations in King County, namely around downtown Seattle where prices are higher on average.