

Step1. Open the data/house.arff Dataset

Click the “*Open file...*” button to open a data set and double click on the “*data*” directory.

Weka provides a number of small common machine learning datasets that you can use to practice on.

Select the “*house.arff*” file to load the house dataset.

The screenshot shows the Weka Explorer window with the 'Preprocess' tab selected. The 'Filter' section is set to 'None'. The 'Current relation' is 'house' with 7 instances and 6 attributes. The 'Attributes' list includes 'houseSize', 'lotSize', 'bedrooms', 'granite', 'bathroom', and 'sellingPrice'. The 'Selected attribute' panel for 'houseSize' shows the following statistics:

Statistic	Value
Minimum	2200
Maximum	4032
Mean	3132
StdDev	655.121

The 'Class' is set to 'sellingPrice (Num)'. A histogram of 'houseSize' is displayed at the bottom right, showing a distribution with a peak around 3116. The status bar at the bottom indicates 'OK' and a 'Log' button.

Step2. Creating the regression model with WEKA

To create the model, click on the **Classify** tab. The first step is to select the model we want to build, so WEKA knows how to work with the data, and how to create the appropriate model:

1. Click the **Choose** button, then expand the **functions** branch
2. Select the **LinearRegression** leaf with using **training set** test option.
3. Click Start to create a model.

Weka Explorer

Preprocess **Classify** Cluster Associate Select attributes Visualize Forecast

Classifier

Choose LinearRegression -S 0 -R 1.0E-8

Test options

Use training set

Supplied test set Set...

Cross-validation Folds 10

Percentage split % 66

More options...

(Num) sellingPrice

Start Stop

Result list (right-click for options)

13:16:20 - functions.LinearRegression

Classifier output

Linear Regression Model

sellingPrice =

-26.6882 * houseSize +

7.0551 * lotSize +

43166.0767 * bedrooms +

42292.0901 * bathroom +

-21661.1208

Time taken to build model: 0.01 seconds

=== Evaluation on training set ===

Time taken to test model on training data: 0 seconds

=== Summary ===

Correlation coefficient	0.9945
Mean absolute error	4053.821
Root mean squared error	4578.4125
Relative absolute error	13.1339 %
Root relative squared error	10.51 %
Total Number of Instances	7

Status OK Log x 0

Step3: Interpreting the regression model

Regression output

$$\text{sellingPrice} = (-26.6882 * \text{houseSize}) + (7.0551 * \text{lotSize}) + (43166.0767 * \text{bedrooms}) + (42292.0901 * \text{bathroom}) - 21661.1208$$

House value using regression model

$$\text{SellingPrice} = (-26.6882 * 2983) + (7.0551 * 9365) + (43166.0767 * 5) + (42292.0901 * 1) - 21661.1208$$

sellingPrice = 222,921

The screenshot shows the Weka Explorer interface with the 'Classify' tab selected. The classifier chosen is 'LinearRegression' with parameters '-S 0 -R 1.0E-8'. The 'Test options' section shows 'Supplied test set' selected. The 'Classifier output' pane displays the following regression equation and evaluation metrics:

```

7.0551 * lotSize +
43166.0767 * bedrooms +
42292.0901 * bathroom +
-21661.1208
Time taken to build model: 0 seconds
=== Predictions on test set ===
inst#   actual   predicted   error
1 230000   222921.571 -7078.429
=== Evaluation on test set ===
Time taken to test model on supplied test set: 0 seconds
=== Summary ===
Correlation coefficient           0
Mean absolute error              7078.429
Root mean squared error          7078.429
Relative absolute error          117.1371 %
Root relative squared error      117.1371 %
Total Number of Instances       1

```

The 'Result list' shows two entries for 'functions.LinearRegression' at different times. The 'Status' bar at the bottom indicates 'OK'.

Step4: Interpret the patterns and conclusions that our model tells us

- **Granite doesn't matter** :It will throw out and ignore columns that don't help in creating a good model. So this regression model is telling us that granite in your kitchen doesn't affect the house's value.
- **Bathrooms do matter**: Since we use a simple 0 or 1 value for an upgraded bathroom, we can use the coefficient from the regression model to determine the value of an upgraded bathroom on the house value. The model tells us it adds \$42,292 to the house value.
- **Bigger houses reduce the value**: Model is telling us that the bigger our house is, the lower the selling price? This can be seen by the negative coefficient in front of the `houseSize` variable. The model is telling us that every additional square foot of the house reduces its

price by \$26? That doesn't make any sense at all. How should we interpret this? The house size, unfortunately, isn't an independent variable because it's related to the bedrooms variable, which makes sense, since bigger houses tend to have more bedrooms. So our model isn't perfect. But we can fix this. On the **Preprocess** tab, you can remove columns from the data set. Remove the **houseSize** column and create another model. How does it affect the price of my house? How does this new model make more sense? (My amended house value: \$215,554).

