Introduction to Linear Regression

May 5, 2020

Let's say we're interested in houses near Belmont. We gather a sample of 20 homes in the area and look at sales prices for those homes.

In our sample, we find that the average sales price was \$482,000.

If we are trying to predict the price of another house in this area, what would our best guess be?

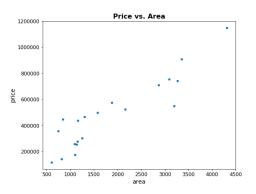
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In the absence of other information, we could go with the average price from our sample, \$482,000, but could we do better if we had more information?

What if we also looked at the square footage of the homes in our sample. Here's a scatterplot of our sample:

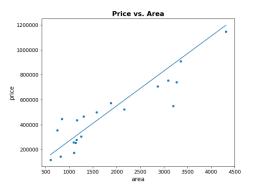


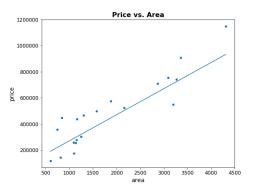
If we know that a house we're interested in is 3200 sqft, could we now make a better guess versus just guessing the average price for the area?

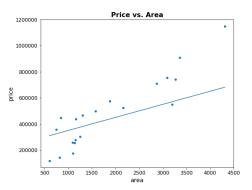
The big idea of least-squares regression is to find a line which describes the relationship between two variables.

Such a line will rarely completely describe the relationship since there will be some uncertainty due either to randomness of due to variables we did not measure.

For example, the price of a home is never completely determined by the square footage, but we can probably make a decent guess about the price by knowing how large the house is.



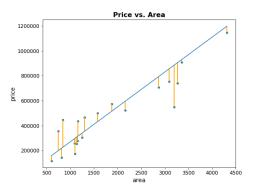


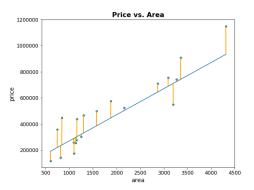


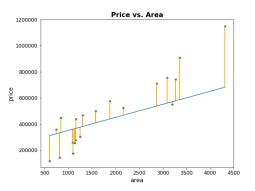
How do we choose the line to use out of all of the options?

We'll choose by looking at the **residuals** - the vertical distance from the line to each point. In this case, the residuals represent the difference between the true price and what we would guess for the price is we used the line to predict price based on square footage.

Because it makes the math work out nicer, we'll really be looking at the squared residuals, but we can get a pretty good idea by looking at the regular residuals.

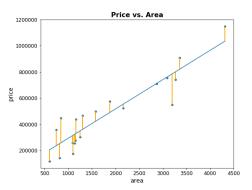






It turns out that the line with the smallest squared residuals is this one, with equation

$$price = 223.7 \cdot (sqft) + 69522.4$$



Least-Squares Regression Line:

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$$50 \cdot \$223.7 = \$11, 185$$

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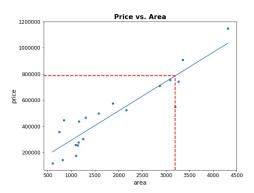
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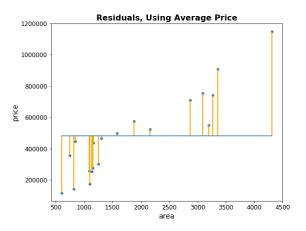
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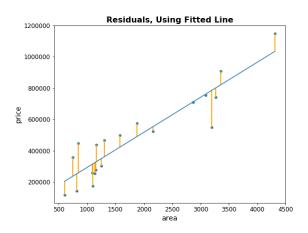
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We can quantify how well our line fits the data by using the coefficient of determination, or R^2 .

We can understand \mathbb{R}^2 by comparing the residuals for our fitted line versus the residuals if we only used the average price to make our predictions.





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In this case, the residuals were reduced by a significant amount. Consequently, the \mathbb{R}^2 value is equal to 0.842.