

# Regression

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September 21, 2021

# Content Questions

## Linear Regression Predictions

| dimension | weight |
|-----------|--------|
| $b$       | 1      |
| $w_1$     | 2.0    |
| $w_2$     | -1.0   |
| $\sigma$  | 1.0    |

1.  $\mathbf{x}_1 = \{0.0, 0.0\}; y_1 =$

2.  $\mathbf{x}_2 = \{1.0, 1.0\}; y_2 =$

3.  $\mathbf{x}_3 = \{.5, 2\}; y_3 =$

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## Probabilities

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| $w_0$     | 1      |
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$$p(y|x) = y \sim N\left(b + \sum_{j=1}^p w_j x_j, \sigma^2\right)$$

$$p(y|x) = \frac{\exp\left\{-\frac{(y-\hat{y})^2}{2}\right\}}{\sqrt{2\pi}}$$

1.  $p(y_1 = 1 | \mathbf{x}_1 = \{0.0, 0.0\}) =$
2.  $p(y_2 = 3 | \mathbf{x}_2 = \{1.0, 1.0\}) =$
3.  $p(y_3 = -1 | \mathbf{x}_3 = \{.5, 2\}) =$

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2.  $p(y_2 = 3 | \mathbf{x}_2 = \{1.0, 1.0\}) = 0.242$
3.  $p(y_3 = -1 | \mathbf{x}_3 = \{.5, 2\}) =$

## Probabilities

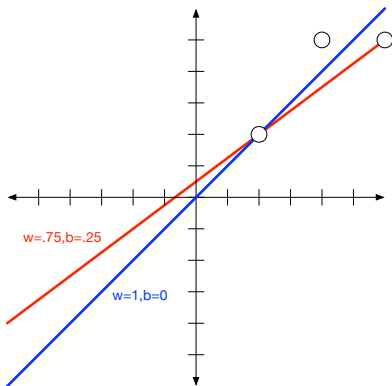
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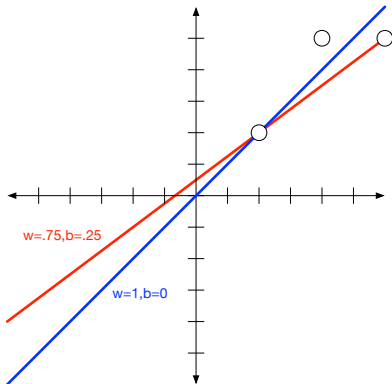
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Consider these points and data

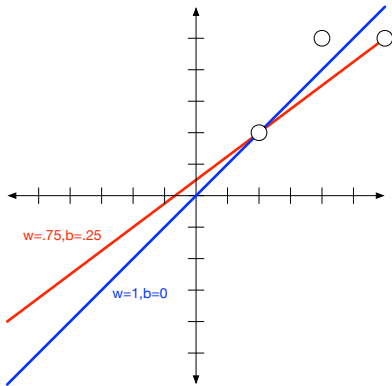


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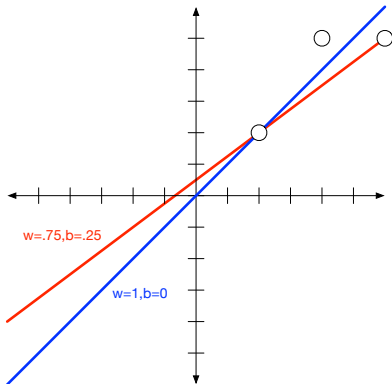
Which is the better OLS solution?

Consider these points and data



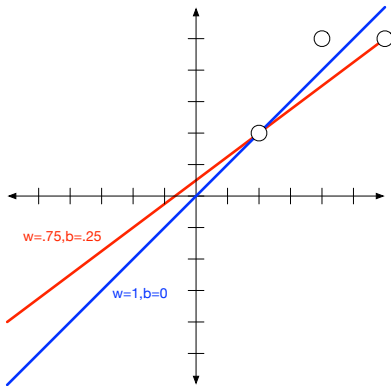
Blue! It has lower RSS.

Consider these points and data



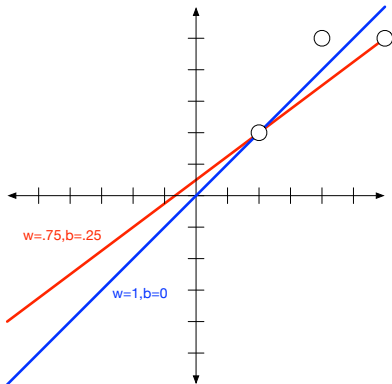
What is the RSS of the better solution?

Consider these points and data



$$\frac{1}{2} \sum_i r_i^2 = \frac{1}{2} ((1-1)^2 + (2.5-2)^2 + (2.5-3)^2) = \frac{1}{4}$$

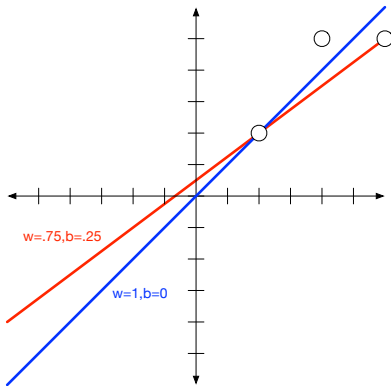
Consider these points and data



What is the RSS of the red line?



Consider these points and data



$$\frac{1}{2} \sum_i r_i^2 = \frac{1}{2} ((1-1)^2 + (2.5-1.75)^2 + (2.5-2.5)^2) = \frac{3}{8}$$