

Errors and residuals in a regression model

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$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_3 X_3 + ... + \beta_k X_k + error$$

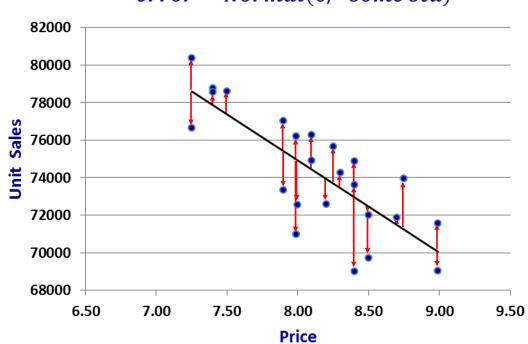
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$$error \sim Normal(0, some std)$$









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Linear Regression for Business Statistics

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Population Mean: $\mu \leftarrow$ Fixed but unknown

Sample Mean :



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Population Mean:
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 Fixed but unknown

Sample Mean:
$$\overline{x} \leftarrow \overline{x} \sim Normal(\mu, some std)$$



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Some important results...

$$\frac{b_0-\beta_0}{s_{b_0}} \sim t_{n-k-1}$$

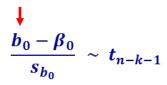
n = number of observations

k = number of "X" variables

n-k-1 = residual degrees of freedom



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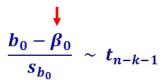
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. . .

These results enable us to...,

. . .

- test stability and precision of coefficients
- conduct hypothesis testing in a regression





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