



Natural Log transformation in a Regression Model.



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Why would we transform variables in a Regression?

Regression is a linear procedure.



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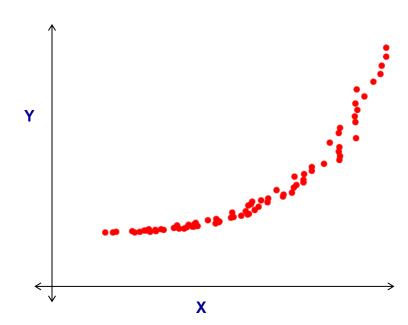
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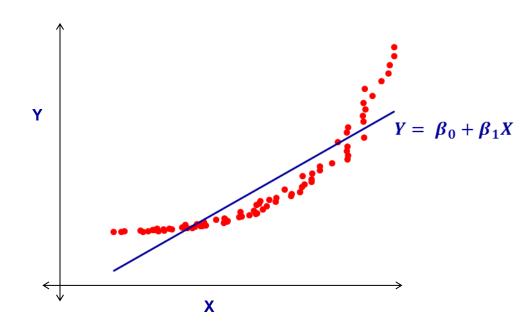
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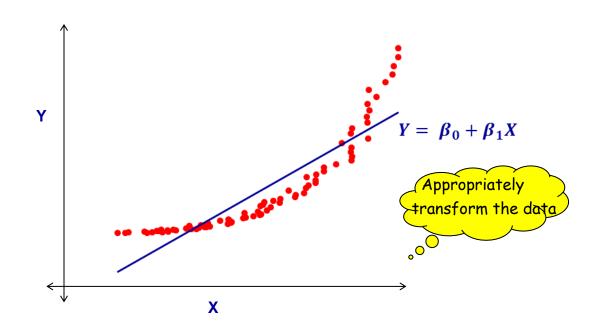
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Log-log Model

Semi-log Model

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The Natural Log function. Some properties...

- □ The symbol we will use for Natural Log is "LN"
- \Box If LN(A) = B, then A = EXP(B), where "EXP" is the exponential function.
- \square LN(A) + LN(B) = LN(A*B)
- \square LN(A) LN(B) = LN(A/B)
- \square LN(1) = 0
- LN(0) is not defined
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The Log-log Model: $LN(Y) = \beta_0 + \beta_1 LN(X_1) + \beta_2 LN(X_2) + \dots$

The Semi-log Model: $LN(Y) = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + ...$



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When the **natural log of** X_1 increases by one unit, then the **natural log of** Y increases by β_1 **units**, all other variables remaining at the same level.

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If the company increases the natural log of promotion expenditure by one unit then the natural log of unit sales of the toy will increase by 1.43 units, all other variables remaining at the same level.

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