

MONASH BUSINESS SCHOOL

Forecasting: principles and practice

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2.2 Transformations

##Variance stabilization

If the data show different variation at different levels of the series, then a transformation can be useful.

Denote original observations as y_1, \ldots, y_n and transformed observations as w_1, \ldots, w_n .

Mathematical transformations for stabilizing variation

Square root
$$w_t = \sqrt{y_t}$$

Cube root $w_t = \sqrt[3]{y_t}$ Increasing

Logarithm $w_t = \log(y_t)$ strength

Logarithms, in particular, are useful because they are more interpretable; changes in a log value are relative (nercent).

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##Box-Cox transformations

Each of these transformations is close to a member of the family of **Box-Cox transformations**:

$$w_t = \begin{cases} \log(y_t), & \lambda = 0; \\ (y_t^{\lambda} - 1)/\lambda, & \lambda \neq 0. \end{cases}$$

- λ = 1: (No substantive transformation)
- $\lambda = \frac{1}{2}$: (Square root plus linear transformation)
- λ = 0: (Natural logarithm)
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