## Choice of the functional form

## What if...





## Exponential function

Exponential functions are functions which can be represented by graphs similar to the graph on the right




Green $=e^{x}$
Black $=3^{x}$
$\operatorname{Red}=2^{x}$

As you could see in the graph, the larger the base, the faster the function increased

If we place a negative sign in front of the x , the graphs will be reflected(flipped) across the $y$-axis


$$
\text { Yellow }=4^{-x} \quad \text { Green }=\mathbf{e}^{-\mathbf{x}}
$$

$$
\text { Black }=3^{-x}
$$

Red $=\mathbf{2}^{-\mathrm{x}}$

## Exponential function

- Exponential functions decrease if $0<\mathrm{b}_{1}<1$ and increase if $b_{1}>1$



## Power function



## Logarithmic function



## Hyperbolic function



## Quadratic function

## Logistic function

## Logistic Growth Function



If growth begins slowly, then increases rapidly and eventually levels off, the data often can be model by an "S-curve", or a logistic function.

## General information

- NONLINEAR MODELS OFTEN ARE USED FOR SITUATION IN WHICH THE RATE OF INCREASE OR DECREASE IN THE DEPENDENT VARIABLE (WHEN PLOTTED AGAINST A PARTICULAR INDEPENDENT VARIABLE) IS NOT CONSTANT.


## General information

- SOME OF THESE MODELS REQUIRED A TRANSFORMATION TO THE INDEPENDENT VARIABLE.


# Transformation 

## - Logarithms <br> - Substitution

Data transformations can be used to convert an equation into a linear form

## Exponential function

| Model | Scatter plot | Transformation | Model after <br> transformation |
| :--- | :--- | :--- | :--- |
| $\hat{y}=b_{0} b_{1}{ }^{x}$ <br> exponential <br> function |  |  | Logarithms |
| $y^{\prime}=b_{0}{ }^{\prime}+b_{1} x^{\prime}$ |  |  |  |
| where $\mathrm{y}^{\prime}=\log \mathrm{y}$ |  |  |  |
| $\mathrm{b}_{0}{ }^{\prime}=\log \mathrm{b}_{0}$ |  |  |  |
| $\mathrm{~b}_{1}=\log \mathrm{b}_{1}$ |  |  |  |

## Power function

| Model | Scatter plot | Transformation | Model after <br> transformation |
| :--- | :--- | :--- | :--- |
| $\hat{y}=b_{0} x^{b 1}$ <br> power <br> function |  | Logarithms | $y^{\prime}=b_{0}{ }^{\prime}+b_{1} x^{\prime}$ <br> where $\mathrm{y}^{\prime}=\log \mathrm{y}$ <br> $\mathrm{x}^{\prime}=\log \mathrm{x}$ <br> $\mathrm{b}_{0}{ }^{\prime}=\log \mathrm{b}_{0}$ |

## Quadratic function

$\left.\begin{array}{|l|c|l|l|}\hline \text { Model } & \text { Scatter plot } & \text { Transformation } & \begin{array}{l}\text { Model after } \\ \text { transformation }\end{array} \\ \hline \begin{array}{l}\hat{y}=b_{0}+b_{1} x+b_{2} x^{2} \\ \text { quadratic function }\end{array} & & / & \text { Substitution }\end{array} \begin{array}{l}y=b_{0}+b_{1} x_{1}+b_{2} x_{2} \\ \text { where } \\ \mathrm{x}_{1}=\mathrm{x} \\ \mathrm{x}_{2}=\mathrm{x}^{2}\end{array}\right]$

## Polynomial function

| Model | Scatter <br> plot | Transformation | Model after <br> transformation |
| :--- | :--- | :--- | :--- |
| $\hat{y}=b_{0}+b_{1} x+b_{2} x^{2}+b_{3} x^{3}+\ldots+b_{k} x^{k}$ <br> polynomial function |  | Substitution | $y=b_{0}+b_{1} x_{1}+b_{2} x_{2}+b_{3} x_{3}+\ldots+b_{k} x_{k}$ <br> where |
|  |  | $\mathrm{x}_{1}=\mathrm{x}$ <br> $\mathrm{x}_{2}=\mathrm{x}^{2}$ <br> $\mathrm{x}_{3}=\mathrm{x}^{3}$ <br> $\mathrm{x}_{\mathrm{k}}=\mathrm{x}^{\mathrm{k}}$ |  |

## Hyperbolic function

| Model | Scatter plot | Transformation | Model after <br> transformation |
| :--- | :--- | :--- | :--- |
| $\hat{y}=b_{0}+b_{1} \frac{1}{x}$ <br> hyperbolic <br> function |  | Substitution | $y=b_{0}+b_{1} x^{\prime}$ <br> where <br> $\mathrm{x}^{\prime}=1 / \mathrm{x}$ |

## Logarithmic function

| Model | Scatter plot | Transformation | Model after <br> transformation |
| :--- | :--- | :--- | :--- |
| $\hat{y}=b_{0}+b_{1} \ln x$ <br> logarithmic <br> function |  | Substitution | $y=b_{0}+b_{1} x^{\prime}$ <br> where <br> $\mathrm{x}^{\prime}=\ln \mathrm{x}$ |

## Logistic function

\(\left.$$
\begin{array}{|l|l|l|l|}\hline \text { Model } & \text { Scatter plot } & \text { Transformation } & \begin{array}{l}\text { Model after } \\
\text { transformation }\end{array} \\
\hline \begin{array}{l}\hat{y}=\frac{b_{0}}{1+b_{1} e^{-x}}\end{array} & & \text { Substitution } & \begin{array}{l}y^{\prime}=b_{0}{ }^{\prime}+b_{1}{ }^{\prime} x^{\prime} \\
\text { where } \\
\text { logistic } \\
\text { function }\end{array}
$$ <br>

\mathrm{y}^{\prime}=1 / \mathrm{y}\end{array}\right\}\)| $\mathrm{x}^{\prime}=\mathrm{e}^{-\mathrm{z}}$ |
| :--- |
| $\mathrm{b}_{0}{ }^{\prime}=1 / \mathrm{b}_{0}$ |
| $\mathrm{~b}_{1}{ }^{\prime}=\mathrm{b}_{1} / \mathrm{b}_{0}$ |

## Linear function

| X | Y | $\Delta \mathrm{X}$ | $\Delta \mathrm{Y}$ | $\Delta \mathrm{Y} / \Delta \mathrm{X}$ |
| :---: | :---: | :---: | :---: | :---: |
| 1,2 | 3,08 | - | - | - |
| 1,4 | 3,58 | 0,2 | 0,5 | 2,50 |
| 1,6 | 4,08 | 0,2 | 0,5 | 2,50 |
| 1,8 | 4,58 | 0,2 | 0,5 | 2,50 |
|  | $\mathrm{~b}_{1}=2,50$ |  |  |  |

## Exponential function

| X | Y | $\Delta \mathrm{X}$ | $\Delta \mathrm{Y}$ | $\Delta \mathrm{Y} / \Delta \mathrm{X}$ | $\log \mathrm{y}$ | $\log \mathrm{y} / \mathrm{x}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1,2 | 4,963 | - | - | - | 0,6957 | 0,580 |
| 1,4 | 6,482 | 0,2 | 1,519 | 7,595 | 0,8117 | 0,580 |
| 1,8 | 11,086 | 0,4 | 4,604 | 11,51 | 1,0448 | 0,580 |
| 3 | 54,872 | 1,2 | 43,786 | 36,4883 | 1,7394 | 0,580 |

$\mathrm{b}_{1}=10^{0,580}=3,8$

## Power function

| X | Y | $\Delta \mathrm{X}$ | $\Delta \mathrm{Y}$ | $\Delta \mathrm{Y} / \Delta \mathrm{X}$ | $\log \mathrm{y}$ | $\log \mathrm{y} / \mathrm{x}$ | $\log \mathrm{x}$ | $\log \mathrm{y} / \log \mathrm{x}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1,4 | 2,027 | - | - | - | 0,3069 | 0,219 | 0,146 | 2,10 |
| 2 | 4,28 | 0,6 | 2,253 | 3,755 | 0,6314 | 0,316 | 0,301 | 2,10 |
| 5 | 29,37 | 3 | 25,085 | 8,36167 | 1,4678 | 0,294 | 0,699 | 2,10 |
| 9 | 100,9 | 4 | 71,535 | 17,88375 | 2,0039 | 0,223 | 0,954 | 2,10 |

$$
b_{1}=2,10
$$

## Comparison

## EXPONENTIAL

POWER

Independent variable is a power exponent

## Form of model:

$$
\hat{y}=b_{0} \cdot b_{1}^{x} \quad \hat{y}=b_{0} \cdot x^{b_{1}}
$$

$\mathrm{b}_{0}$ - is the value of Y if independent variable is equal to zero.
$\mathrm{b}_{1}$ - is the growth rate Y. If the independent variable increases 1 unit, the dependent variable will change (increase, if $\mathrm{b}_{1}>1$, or decrease, if $\left.\mathrm{b}_{1}<1\right) \mathrm{b}_{1}$ times, on average $\{$ or $\left(b_{1}-1\right) \times 100[\%]$, on average $\}$.
$\mathrm{b}_{0}-$ is the value of Y if independent variable is equal to one
$\mathrm{b}_{1}$ - is the elasticity Y. If the independent variable increases $1 \%$, the dependent variable will change (increase, if $\mathrm{b}_{1}>0$, or decrease, if $\left.\mathrm{b}_{1}<0\right) \mathrm{b}_{1} \%$, on average.

Independent variable is a power base

## Interpretation of the coefficients

## Comparison

## Linear transformation - logarithms

$$
\log \hat{y}=\log b_{0}+x \cdot \log b_{1} \quad \underline{\text { Linear form }} \quad \log \hat{y}=\log b_{0}+b_{1} \cdot \log x
$$

## Parameters estimation - OLS:



Matrix and vector:

$$
\begin{array}{ll}
X^{T} X=\left[\begin{array}{cc}
n & \sum x \\
\sum x & \sum x^{2}
\end{array}\right] & \log X^{T} \log X=\left[\begin{array}{cc}
n & \sum \log x \\
\sum \log x & \sum(\log x)^{2}
\end{array}\right] \\
X^{T} \log Y=\left[\begin{array}{c}
\sum \log y \\
\sum x \cdot \log y
\end{array}\right] & \log X^{T} \log Y=\left[\begin{array}{c}
\sum \log y \\
\sum \log x \cdot \log y
\end{array}\right]
\end{array}
$$

## Comparison

After $\log b_{0}$ and $\log b_{1}$ are estimated we should check goodness of fit (standard error of the estimate, indetermination coefficient, determination coefficient, test parameters individually and check residuals' characteristics - at least linearity) for the linear form.

$$
\begin{array}{cc}
\log \hat{y}=\log b_{0}+x \cdot \log b_{1} & \log \hat{y}=\log b_{0}+b_{1} \cdot \log x \\
S_{e}=\sqrt{\frac{\sum\left(\log y_{i}-\log \hat{y}_{i}\right)^{2}}{n-k-1}} & S_{e}=\sqrt{\frac{\sum\left(\log y_{i}-\log \hat{y}_{i}\right)^{2}}{n-k-1}}
\end{array}
$$

To interpret the results, antilog $b_{0}$ and $b_{1}$ should be calculated

To interpret the results, antilog $b_{0}$ should be calculated

