

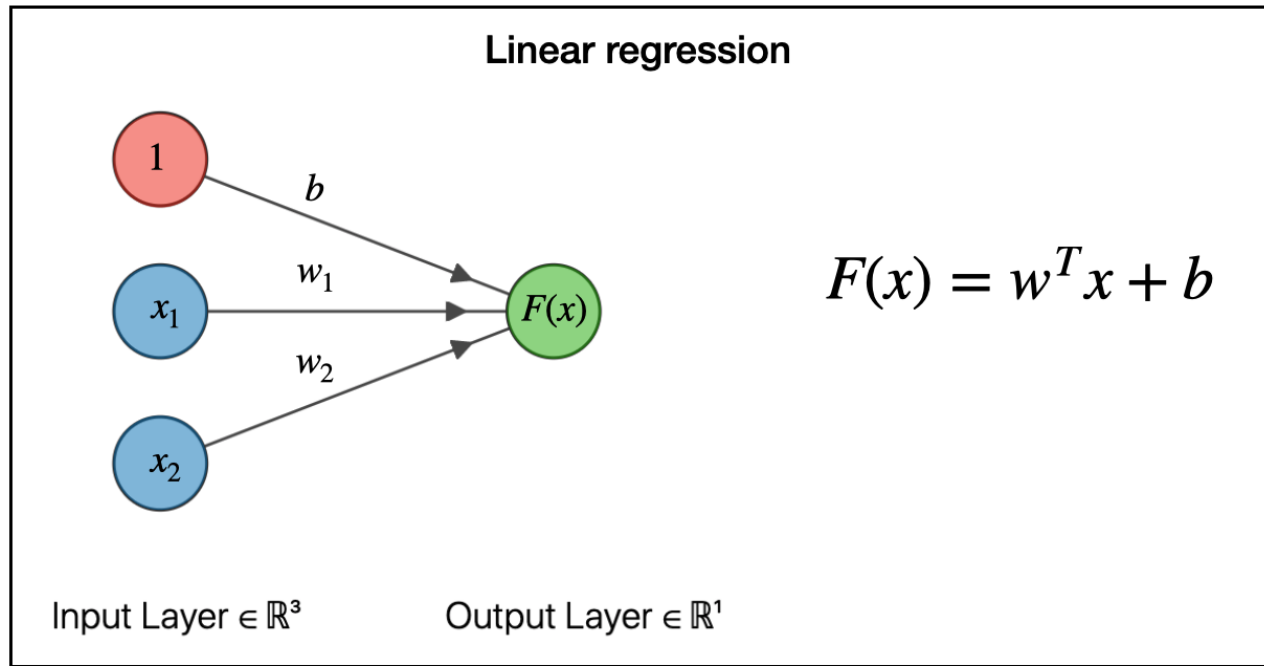
SNN Example 15 - Linear Models

The major reason for the popularity of neural networks is their ability to model nonlinear problems; i.e., classification problems that cannot be solved simply by drawing a single hyperplane between classes, and regression problems that cannot be solved simply by drawing a hyperplane through the data.

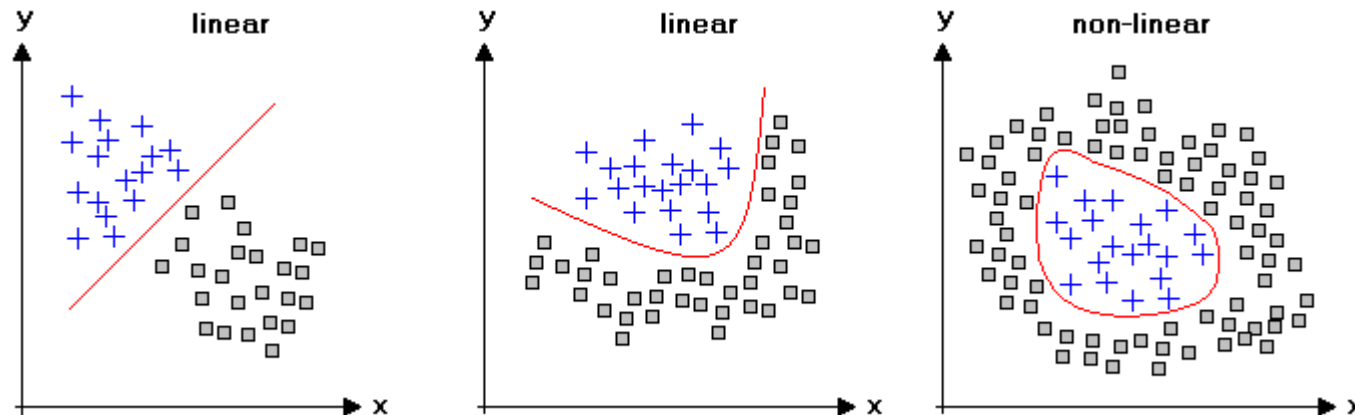
Despite this, **linear models should not be neglected**. It is not too uncommon to find that a problem that was perceived to be difficult and nonlinear can actually be solved adequately by linear techniques, and a linear model always provides a good benchmark against which to judge more complex techniques.

STATISTICA Neural Networks builds linear models using a special linear network type. A linear network has only two layers: an input layer, and an output layer with linear [SynapticP](#) and [activation](#) functions. See also, [Synaptic Functions](#) and [Activation Functions](#).

Linear networks, like the output layer of an [RBF](#) network, can be optimized directly using the [pseudo-inverse](#) technique. They are designed principally for regression problems, but are equally able to perform a simple discriminant analysis on classification problems. See also, [Linear Networks](#) and [Radial Basis Function Networks](#).



$$F(x) = w_1 \cdot x_1 + w_2 \cdot x_2 + 1 \cdot b$$



To create a linear model of the Iris regression problem, use the [Custom Network Designer](#) dialog as usual, setting the *Network type* option to *Linear*. Linear networks do not have hidden units, so there are no further settings to make.

Training a Linear network is equally straightforward - just press the *OK* button on the *Train Linear Network* dialog.

Linear models are actually ill-suited to the Iris problem, as the *VERSICOL* and *VIRGINIC* species are not linearly separable, and if you generate *Descriptive statistics* you will see a significant amount of misclassification between these two classes.

Custom Network Designer: Irisdat

Dependent: IRISTYPE
Independent: SEPALLEN-PETALWID

Quick | Time Series

Network type

- Multilayer Percep
- Radial Basis Fun
- Probabilistic Neu
- Generalized Reg
- Self Organizing P
- Linear
- Principal Components Network
- Clustering Network

Train Linear Network: Irisdat

Quick | Pruning | Classification

Press OK to optimize by Singular Value Decomposition

Results (Run Models): Irisdat

Index	Profile	Train Perf.	Select Perf.	Test Perf.	Train Error
1	Linear 4:4-3:1	0,855263	0,837838	0,810811	0,288734

Quick | Advanced | Predictions | Sensitivity | Descriptive Statistics

- User defined case
- Response graph
- Response surface
- ROC curve
- PCA eigenvalues
- Topological map
- Network illustration
- Time series projection
- Training graph

Generate spreadsheet of eigenvalues (applicable only if the last trained network was a PCA)

Model Summary Report (Irisdat)

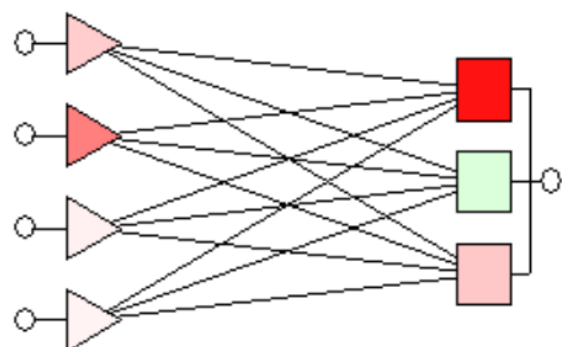
Index	Profile	Train Perf.	Select Perf.	Test Perf.	Train Error	Select Error	Test Error	Training/Members	Note	Inputs	Hidden(1)	Hidden(2)
1	Linear 4:4-3:1	0,855263	0,837838	0,810811	0,288734	0,301891	0,322531		PI	4	0	0

	Confusion Matrix - IRISTYPE(1) (Irisdat)		
	SETOSA	VIRGINIC	VERSICOL
SETOSA.1	50,0000	0,0000	0,0000
VIRGINIC.1	0,0000	42,0000	16,0000
VERSICOL.1	0,0000	8,0000	34,0000

	Classification (1) (Irisdat)		
	IRISTYPE.SETOSA.1	IRISTYPE.VIRGINIC.1	IRISTYPE.VERSICOL.1
Total	50,0000	50,0000	50,0000
Correct	50,0000	42,0000	34,0000
Wrong	0,0000	8,0000	16,0000
Unknown	0,0000	0,0000	0,0000
Correct(%)	100,0000	84,0000	68,0000
Wrong(%)	0,0000	16,0000	32,0000
Unknown(%)	0,0000	0,0000	0,0000

Profile : Linear 4:4-3:1 , Index = 1

Train Perf. = 0,842105 , Select Perf. = 0,864865 , Test Perf. = 0,837838



	Sensitivity Analysis - 1 (Irisdat)			
	SEPALLEN	SEPALWID	PETALLEN	PETALWID
Ratio.1	1,040763	1,056140	1,860177	1,719268
Rank.1	4,000000	3,000000	1,000000	2,000000