



HAL
open science

Use of PLS Path Modelling to estimate the European Consumer Satisfaction Index (ECSI) model

Marie-Paule Bayol, Anne de La Foye, Michel Tenenhaus, Carole Tellier

► To cite this version:

Marie-Paule Bayol, Anne de La Foye, Michel Tenenhaus, Carole Tellier. Use of PLS Path Modelling to estimate the European Consumer Satisfaction Index (ECSI) model. *Statistica Applicata - Italian Journal of Applied Statistics*, 2000. hal-03110080

HAL Id: hal-03110080

<https://hal.inrae.fr/hal-03110080>

Submitted on 14 Jan 2021

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

Use of PLS Path Modelling to estimate the European Consumer Satisfaction Index (ECSI) model

Marie-Paule Bayol¹, Anne de la Foye¹, Carole Tellier¹ and Michel Tenenhaus²
(1) Cisia-Ceresta, Montreuil, (2) HEC Graduate School of Management, Jouy-en-Josas

Abstract

The European Consumer Satisfaction Index (ECSI) is an economic indicator that measures customer satisfaction. It is an adaptation of the Swedish Customer Satisfaction Barometer and is compatible with the American Customer Satisfaction Index. In this paper the ECSI model is presented in details. The PLS approach used to estimate the model parameters is described. Finally, an example is discussed.

I. The ECSI model

The European Consumer Satisfaction Index (ECSI) is an economic indicator that measures customer satisfaction. It is an adaptation of the Swedish Customer Satisfaction Barometer (Fornell, 1992) and is compatible with the American Customer Satisfaction Index. A model has been derived specifically for the ECSI. In this model, seven interrelated latent variables are introduced. It is based on well-established theories and approaches in customer behaviour and it is to be applicable for a number of different industries.

The ECSI model (figure 1) contains :

1. A core model, i.e. the traditional latent variables : **perceived quality, expectations, perceived value, satisfaction index and loyalty**, shown in bold for the constructs and impacts in solid lines.
2. Two optional latent variables that can be added by national committees : *image and complaints* - shown in italic and impacts in dotted lines.

Other impacts, than the ones shown on figure 1, can be tested.

The variables on the left-hand side are to be seen as drivers for explaining the Customer Satisfaction Index (CSI) and the right hand performance indicator (*loyalty/complaints*). Main causal relationships are indicated.

A set of manifest (observable or measurable) variables is associated with each of the latent variables. This structure is called the ECSI model. The entire model is important for determining the main goal variable, being CSI.

The *perceived quality* concept includes two parts (“software” and “hardware”). With the “hardware” component is meant the quality of the product as such (in the eyes of the customer), while “software” relates to associated service like guarantees given, after sale

service provision, conditions of product display and assortment, documentation and descriptions, etc.

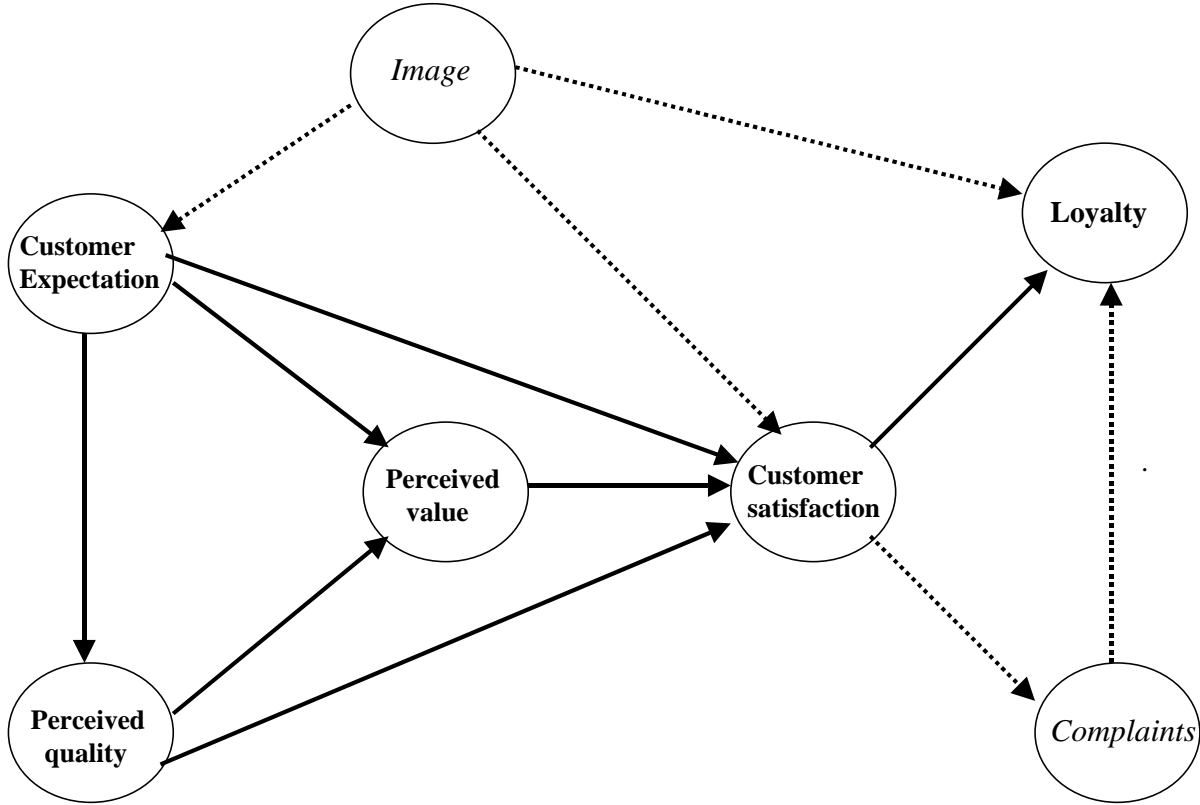


Figure 1: Causality model describing causes and consequences of Customer Satisfaction

Customer expectations relate to the prior anticipations of the said product in the eyes of the customer. Such expectations are the result of active company/product promotion as well as hearsay and prior experience from the product/provider.

Image is an optional variable that relates to the brand name and what kind of associations the customers get from the product/brand/company. The arrows between image and CSI and loyalty respectively could be in reality two-directional.

Perceived value concerns the “value-for-money” aspect as the customer experiences it. It is here seen to be affected by perceived quality as well as by expectations.

Complaints is another optional variable and relates to the intensity of complaints by the clients and the way the company handles these complaints. It is linked (probably with a bi-directional arrow) to CSI and Loyalty.

In table 1 the manifest variables V_{jh} describing the latent variables ξ_j are given for the Mobile Phone Industry.

Table 1 : Measurement Instrument for the Mobile Phone Industry

All the items are scaled from 1 to 10. Scale 1 expresses a very negative point of view on the product and scale 10 a very positive opinion.

| Latent variables | Manifest variables |
|--|---|
| Image (ξ_1) | <ul style="list-style-type: none"> a) It can be trusted in what it says and does b) It is stable and firmly established c) It has a social contribution for the society d) It is concerned with customers e) It is innovative and forward looking |
| Customer Expectations of the overall quality (ξ_2) | <ul style="list-style-type: none"> a) Expectations for the overall quality of “your mobile phone provider” at the moment you became customer of this provider b) Expectations for “your mobile phone provider” to provide products and services to meet your personal need c) How often did you expect that things could go wrong at “your mobile phone provider” |
| Perceived Quality (ξ_3) | <ul style="list-style-type: none"> a) Overall perceived quality b) Technical quality of the network c) Customer service and personal advice offered d) Quality of the services you use e) Range of services and products offered f) Reliability and accuracy of the products and services provided g) Clarity and transparency of information provided |
| Perceived Value (ξ_4) | <ul style="list-style-type: none"> a) Given the quality of the products and services offered by “your mobile phone provider” how would you rate the fees and prices that you pay for them? b) Given the fees and prices that you pay for “your mobile phone provider” how would you rate the quality of the products and services offered by “your mobile phone provider”? |
| Customer Satisfaction (ξ_5) | <ul style="list-style-type: none"> a) Overall satisfaction b) Fulfilment of expectations c) How well do you think “your mobile phone provider” compares with your ideal mobile phone provider ? |
| Customer Complaints (ξ_6) | <ul style="list-style-type: none"> a) You complained about “your mobile phone provider” last year. How well, or poorly, was your most recent complaint handled or b) You did not complain about “your mobile phone provider” last year. Imagine you have to complain to “your mobile phone provider” because of a bad quality of service or product. To what extent do you think that “your mobile phone provider” will care about your complaint? |
| Customer Loyalty (ξ_7) | <ul style="list-style-type: none"> a) If you would need to choose a new mobile phone provider how likely is it that you would choose “your provider” again? b) Let us now suppose that other mobile phone providers decide to lower their fees and prices, but “your mobile phone provider” stays at the same level as today. At which level of difference (in %) would you choose another mobile phone provider? c) If a friend or colleague asks you for advice, how likely is it that you would recommend “your mobile phone provider”? |

These manifest variables V_{jh} are normalised as follows:

The original items V_{jh} , scaled from 1 to 10, are transformed into new normalised variables $x_{jh} = \frac{100}{9}(V_{jh} - 1)$. The minimum possible value of x_{jh} is 0 and its maximum possible value is equal to 100. If there are missing data for variable x_{jh} , they are replaced by the mean \bar{x}_{jh} of this variable.

Relation between the manifest variables and the latent variables

Each latent variable ξ_j is indirectly observable by a set of manifest variables x_{jh} . Each manifest variable is related to its latent variable by simple regression:

$$(1) \quad x_{jh} = \lambda_{jh0} + \lambda_{jh}\xi_j + \varepsilon_{jh}$$

where ξ_j has mean m_j and standard deviation 1. It's a reflexive scheme: each manifest variables x_{jh} reflects its latent variable ξ_j . The usual hypotheses on the residuals are made.

The normalisation of the latent variable chosen by Wold (1985) – ξ_j has a standard deviation equal to one – is arbitrary. Another normalisation has been proposed by Fornell (1992), but both Wold and Fornell latent variables are co-linear. In other words, the Fornell latent variable can be deduced from the Wold latent variable by a linear relation.

Relation between the latent variables

The causality model described in figure 1 leads to linear equations relating the latent variables (structural equation modelling):

$$(2) \quad \xi_j = \beta_{j0} + \sum_i \beta_{ji}\xi_i + \zeta_j$$

Let's write the six structural equations corresponding to figure 1:

- (2.1) Customer Expectation = $\beta_{20} + \beta_{21}\text{Image} + \zeta_2$
- (2.2) Perceived Quality = $\beta_{30} + \beta_{32}\text{Customer Expectation} + \zeta_3$
- (2.3) Perceived Value = $\beta_{40} + \beta_{42}\text{Customer Expectation} + \beta_{43}\text{Perceived Quality} + \zeta_4$
- (2.4) ECSI = $\beta_{50} + \beta_{51}\text{Image} + \beta_{52}\text{Customer Expectation} + \beta_{53}\text{Perceived Quality} + \beta_{54}\text{Perceived Value} + \zeta_5$
- (2.5) Customer Complaint = $\beta_{60} + \beta_{65}\text{ECSI} + \zeta_6$
- (2.6) Customer Loyalty = $\beta_{70} + \beta_{71}\text{Image} + \beta_{75}\text{ECSI} + \beta_{76}\text{Customer Complaint} + \zeta_7$

The usual hypotheses on the residuals are made.

II. Partial Least Square (PLS) Estimation of the ECSI model

PLS Path Modelling of Herman Wold (Wold, 1985, Lohmöller, 1989, Fornell & Cha, 1994, Tenenhaus, 1999) is used to estimate the ECSI model parameters. We remind in this paper the various steps of the PLS algorithm and describe the specific options chosen for the estimation

of the ECSI model. Computation is carried out using the LVPLS 1.8 software of J.B. Lohmöller (1987).

First of all, in the ECSI model, PLS is applied to the raw manifest variables x_{jh} . They are not standardised.

Estimation of the latent variables

The latent variables ξ_j are estimated according to the following procedure.

External estimation Y_j of the standardised latent variable ($\xi_j - m_j$)

The standardised latent variables (mean = 0 and standard deviation = 1) are estimated as linear combinations of their centred manifest variables:

$$(3) \quad Y_j \propto [\sum w_{jh} (x_{jh} - \bar{x}_{jh})]$$

where the symbol “ \propto ” means that the left variable represents the standardised right variable. The standardised latent variable is finally written as

$$(4) \quad Y_j = \sum \tilde{w}_{jh} (x_{jh} - \bar{x}_{jh})$$

The mean m_j is estimated by $\hat{m}_j = \sum \tilde{w}_{jh} \bar{x}_{jh}$ and the latent variable ξ_j by $\sum \tilde{w}_{jh} x_{jh} = Y_j + \hat{m}_j$.

Internal estimation Z_j of the standardised latent variable ($\xi_j - m_j$)

Following Wold's (1985) original PLS algorithm, the centroid scheme is used:

$$(5) \quad Z_j \propto \sum_{i: \xi_i \text{ is connected to } \xi_j} e_{ji} Y_i$$

where e_{ji} is the sign of the correlation between Y_j and Y_i . Two latent variables are connected if there exists a link between the two variables: an arrow goes from one variable to the other in the arrow diagram describing the causality model.

Estimation of the weight w_{jh}

Weights w_{jh} are estimated using the *Mode A* (or *Outward Mode*) way of calculation. The weight w_{jh} is the covariance between the manifest variable x_{jh} and the internal estimation Z_j :

$$(6) \quad w_{jh} = \text{cov}(x_{jh}, Z_j)$$

The PLS algorithm consists in beginning with an arbitrary choice of weights w_{jh} , for example w_{j1} is fixed to 1 and all the other w_{jh} to 0. Then steps 3, 5 and 6 are iterated until convergence (not guaranteed, but practically always encountered in practice).

Specific calculation of the latent variables for the ECSI model

In the ECSI model, following Fornell (1992), each latent variable is obtained as a weighted average of its manifest variables:

$$(7) \quad \hat{\xi}_j = \frac{\sum_h \tilde{w}_{jh} x_{jh}}{\sum_h \tilde{w}_{jh}}$$

In this operation it is supposed that all the normalised weights ($\tilde{w}_{jh} / \sum_h \tilde{w}_{jh}$) are positive. If some of these normalised weights are negative, the corresponding variable x_{jh} should be removed from the model as this variable does not correctly describe its latent variable.

These constructs have more practical meaning than the standardised latent variables as they are in the same units than the manifest variables scaled from 0 to 100. The linear relationship between the Wold's latent variable Y_j and the Fornell's latent variable $\hat{\xi}_j$ is clear from equations (4) and (7):

$$(8) \quad \hat{\xi}_j = \frac{1}{\sum_h \tilde{w}_{jh}} (Y_j + \hat{m}_j)$$

Estimation of the structural equation

The structural equations (2) are estimated by individual multiple regressions where the latent variables ξ_j are replaced by their estimations Y_j . These regressions are standard outputs of the Lohmöller's program. The significance levels of the regression coefficients are calculated using the usual Student's t statistic. We prefer this solution to the Jackknife solution for two reasons: (1) The Jackknife solution is not available in the Lohmöller's program when the number of subjects is too high (more than 400), and (2) When the Jackknife solution is available in the Lohmöller's program (less than 400 subjects) the Student's t and the Jackknife solutions are quite comparable.

III. Construction of the Consumer Satisfaction Index for a mobile phone provider

We will study in this section how to use the Lohmöller's program to compute the Customer Satisfaction Index. The data represent the answers of 250 consumers of a mobile phone provider in a specific European country. The program code for the Lohmöller's program LVPLS 1.8 is given in annex 1. The results are given in annex 2.

Estimation of the latent variables

The Lohmöller's program gives the standardised latent variables $Y_j = \sum \tilde{w}_{jh} (x_{jh} - \bar{x}_{jh})$. The weights \tilde{w}_{jh} appear in the program output in column *Weight* of table *Outer model* and

the estimated mean $\hat{m}_j = \sum \tilde{w}_{jh} \bar{x}_{jh}$ in the column *Mean* of table *Inner model*. For example the standardised latent variable CSI_std is calculated according to the following formula:

$$CSI_std = 0.0158 \times C_sat1 + 0.0231 \times C_sat2 + 0.0264 \times C_sat3 - 4.6523$$

For the ECSI model the latent variables are calculated as

$$\hat{\xi}_j = \frac{\sum_h \tilde{w}_{jh} X_{jh}}{\sum_h \tilde{w}_{jh}}$$

For example the latent variable “Customer Satisfaction” is calculated as

$$CSI = \frac{0.0158 \times C_sat1 + 0.0231 \times C_sat2 + 0.0264 \times C_sat3}{0.0158 + 0.0231 + 0.0264}$$

This is a weighted average of the manifest variables C_sat1 to C_sat3 scaled from 0 to 100.

The means and standard deviations of the various latent variables are given in table 2:

Table 2 : Mean and standard deviation of the latent variables

| | N | Minimum | Maximum | Mean | Std. Deviation |
|-----------------------|-----|---------|---------|---------|----------------|
| IMAGE | 250 | 26.49 | 100.00 | 72.6878 | 13.7660 |
| CUSTOMER EXPECTATION | 250 | 25.85 | 100.00 | 72.3198 | 14.1259 |
| PERCEIVED QUALITY | 250 | 23.95 | 100.00 | 74.5765 | 14.2573 |
| PERCEIVED VALUE | 250 | .00 | 100.00 | 61.5887 | 20.5987 |
| CUSTOMER SATISFACTION | 250 | 23.68 | 100.00 | 71.2876 | 15.3417 |
| COMPLAINT | 250 | .00 | 100.00 | 67.4704 | 25.2684 |
| LOYALTY | 250 | 1.29 | 100.00 | 69.1757 | 21.2668 |

In table 3, we check that each manifest variable is more correlated to its own latent variable than to the other latent variables. To make this table easier to read, correlations below 0.5 are not shown. You may notice that the manifest variable Loyalty2 does not correctly describe its latent variable (in fact $\text{cor}(\text{Loyalty2}, \text{Loyalty}) = 0.272$). This variable should be removed from the model. In fact it is difficult to give a meaningful answer to this item.

The ECSI model for a Mobile Phone Provider

The causality model of figure 2 summarises the various structural regressions of the ECSI model. The path coefficients are the standardised regression coefficients. The R^2 are also shown. These coefficients appear in Annex 2 in tables *Path coefficients* and *Inner model*. As the weights for variables Perceived value and Customer Loyalty are negative, we have to take the opposite of the path coefficients related to these two variables. The significance levels shown next to the path coefficients have been calculated using simple or multiple regressions. The significant arrows are in bold.

Variables Image, Perceived value and quality have a significant impact on Customer Satisfaction. However the most important impact on Customer Satisfaction is Perceived quality (.544). Image and Perceived value have less impact (.200 and .153). It is not surprising that actual qualities of the mobile phone provider are much more important for the customer than some abstract marketing characteristics. Customer Expectation has no direct impact on Customer Satisfaction. Loyalty is a very important factor in the mobile phone industry. It mainly depends upon Customer Satisfaction (.466) and to a less extent Image (.212). It is interesting to note that Complaints depends on Customer Satisfaction, but has no direct impact on Loyalty. Of course we have to be careful for the interpretation of non significant path coefficients as it can come from a multicollinearity problem. This suggests to use PLS regression (Martens & Næs, 1989, Tenenhaus, 1998) instead of multiple regression.

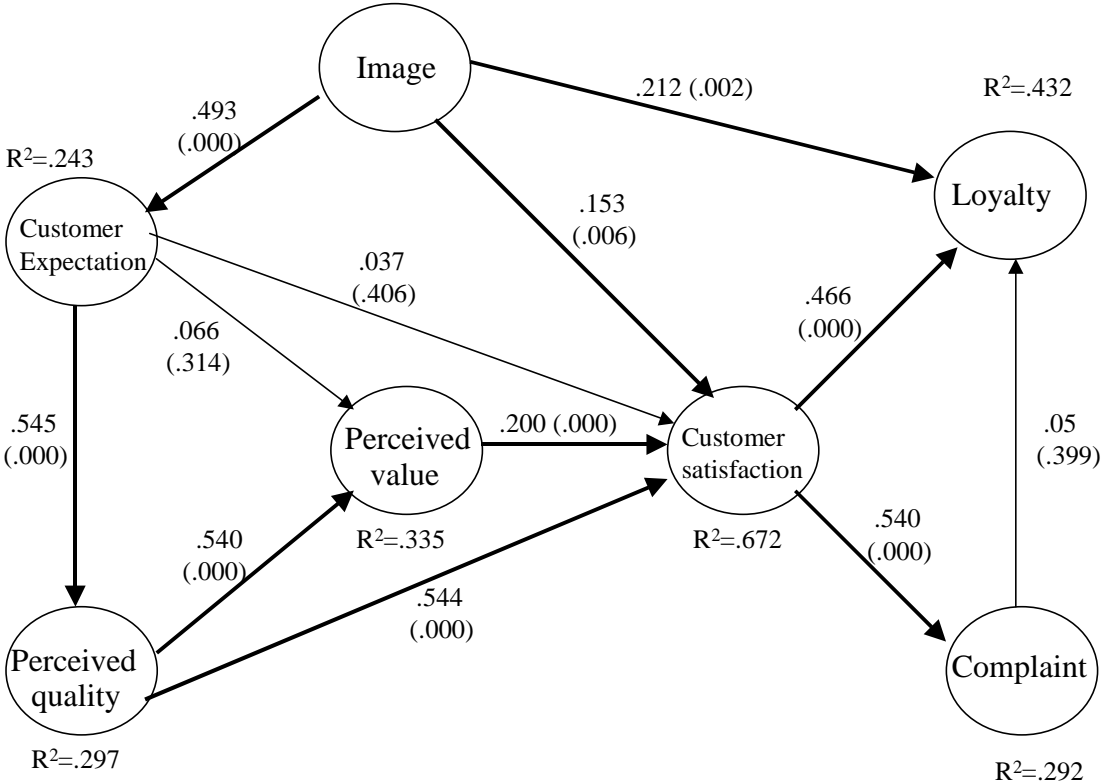


Figure 2 : ECSI Causality model for a mobile telephone provider

Table 3 : Correlations between manifest variables and latent variables

| | Image | Customer expectation | Perceived quality | Perceived value | Customer satisfaction | Complaint | Loyalty |
|-----------|-------|----------------------|-------------------|-----------------|-----------------------|-----------|---------|
| Image1 | .717 | | .571 | | .539 | | |
| Image2 | .565 | | | | | | |
| Image3 | .657 | | | | | | |
| Image4 | .791 | | .571 | | .543 | | |
| Image5 | .698 | | .544 | | .500 | | |
| C_exp1 | | .689 | | | | | |
| C_exp2 | | .644 | | | | | |
| C_exp3 | | .724 | | | | | |
| P_qual1 | .622 | .537 | .778 | | .661 | | |
| P_qual2 | . | | .651 | | | | |
| P_qual3 | .621 | | .801 | | .651 | | |
| P_qual4 | . | | .760 | | .587 | | |
| P_qual5 | .599 | | .732 | | .516 | | |
| P_qual6 | .551 | | .766 | | .539 | | |
| P_qual7 | .596 | | .803 | .547 | .707 | | |
| P_val1 | | | | .933 | | | |
| P_val2 | .541 | | .594 | .911 | .631 | | .524 |
| C_sat1 | .558 | | .638 | | .711 | | |
| C_sat2 | .524 | | .672 | | .872 | | |
| C_sat3 | .613 | | .684 | .588 | .884 | .547 | .610 |
| Complaint | | | .537 | | .540 | 1 | |
| Loyalty1 | | | | | | | .854 |
| Loyalty2 | | | | | | | |
| Loyalty3 | .528 | | .537 | | .659 | | .869 |

CONCLUSION

In his 92 paper, Fornell described in great details the marketing foundations of the Swedish Customer Satisfaction Barometer, but he only touched the statistical aspects of the problem. The objective of this paper was to describe very precisely all the Fornell's statistical methodology. We have shown how to use the Lohmöller's program LVPLS 1.8 to estimate the ECSI model. It has been checked on a pilot study that the Fornell's program (the program is not available, but we had access to the results of the pilot study) and the Lohmöller's program give exactly the same results.

REFERENCES

- FORNELL C. (1992): "A National Customer Satisfaction Barometer: The Swedish Experience", *Journal of Marketing*, Vol. 56, 6-21.
- FORNELL C. & CHA J. (1994): "Partial Least Squares", in *Advanced Methods of Marketing Research*, R.P. Bagozzi (Ed.), Basil Blackwell, Cambridge, MA., pp. 52-78.
- LOHMÖLLER J.-B. (1987): *LVPLS Program Manual, Version 1.8*, Zentralarchiv für Empirische Sozialforschung, Köln.
- LOHMÖLLER J.-B. (1989): *Latent Variables Path Modeling with Partial Least Squares*, Physica-Verlag, Heidelberg.
- MARTENS H. & NÆS T. (1989): *Multivariate Calibration*. John Wiley & Sons, New York.
- TENENHAUS M. (1998): *La Régression PLS*. Éditions Technip, Paris
- TENENHAUS M. (1999): "L'approche PLS", *Revue de Statistique Appliquée*, vol. 47, n°2, pp. 5-40.
- WOLD H. (1985): "Partial Least Squares", in *Encyclopedia of Statistical Sciences*, vol. 6, Kotz, S & Johnson, N.L. (Eds), John Wiley & Sons, New York, pp. 581-591.

ANNEX 1

The program code for the Lohmöller LVPLS 1.8 program

```
LVPX
Etude Téléphonie mobile
  7 250 13256 2 100 5 4 0
  5 3 7 2 3 1 3
  0 0 0 0 0 0 0
  0 0 0 0 0 0 0
IMAG1  IMAG2  IMAG3  IMAG4  IMAG5  CUEX1  CUEX2  CUEX3  PERQ1
PERQ2  PERQ3  PERQ4  PERQ5  PERQ6  PERQ7  PERV1  PERV2  CUSA1
CUSA2  CUSA3  CUSCO  CUSL1  CUSL2  CUSL3
  0 111      (2A4,7F2.0)
IMAGE  0 0 0 0 0 0 0
CUS_EXP 1 0 0 0 0 0 0
PER_QUAL 0 1 0 0 0 0 0
PER_VAL  0 1 1 0 0 0 0
ECSI    1 1 1 1 0 0 0
CUS_COMP 0 0 0 0 1 0 0
CUS_LOY 1 0 0 0 1 1 0
  0 0      (2A4,23F7.0,F6.0)
100293 66.67 44.44 44.44 44.44 44.44 33.33 66.67 66.67 55.56 66.67
55.56 33.33 66.67 55.56 44.44 44.44 11.11 22.22 55.56 33.33 66.67
66.67 55.56 44.44 55.56
100382 100.00 88.89 100.00 100.00 100.00 88.89 100.00 100.00 88.89 100.00
88.89 100.00 100.00 88.89 100.00 100.00 100.00 100.00 100.00 100.00 77.78
100.00 100.00 11.11 100.0
100386 77.78 66.67 55.56 33.33 66.67 66.67 66.67 66.67 66.67 66.67
77.78 44.44 66.67 77.78 66.67 66.67 66.67 66.67 77.78 66.67 66.67
55.56 55.56 11.11 66.67
.
.
.
302284 66.67 77.78 44.44 77.78 77.78 77.78 77.78 66.67 77.78
66.67 55.56 77.78 88.89 66.67 88.89 77.78 77.78 77.78 66.67 77.78
77.78 66.67 100.00 77.78
302291 77.78 77.78 55.56 77.78 66.67 66.67 55.56 66.67 66.67
55.56 55.56 77.78 66.67 55.56 66.67 44.44 44.44 77.78 44.44 66.67
11.11 66.67 66.67 66.67
300589 77.78 100.00 77.78 88.89 88.89 100.00 77.78 44.44 88.89
100.00 100.00 77.78 77.78 77.78 88.89 44.44 88.89 66.67 77.78
77.78 88.89 100.00 22.22 100.00
STOP
```

Comments

Lines 3 to 6 of this program describe the specific selected options. They are explained in the program output (annex 2).

Lines 7 to 9 give the names of the manifest variables and their order in the data file.

Line 10 gives the reading format for the structural equations.

Lines 11 to 17 give the structural equations model. When the model is recursive (no loop) the matrix is lower diagonal. This is the case here.

Line 18 gives the reading format for the data.

Next lines contain the data (customer identification and manifest variables x_{jh} (scaled between 0 and 100)).

ANNEX 2

The results

JBL 1.8

=====
-- P L S X --

-- LATENT VARIABLES PATH ANALYSIS --
- PARTIAL LEAST-SQUARES ESTIMATION -

Etude Téléphonie mobile

===== **COMMENTS**

Number of Blocks NBLOCS = 7
Number of Cases NCASES = 250
Number of Dimensions NDIM = 1
Output Quantity OUT = 3256
Inner Weighting Scheme IWGHT = 2
Number of Iterations NITER = 100
Estimation Accuracy EPS = 5
Analysed Data Metric METRIC = 4

Centroid Scheme

Manifest variables are not standardised

=====
Block N-MV Deflate LV-Mode Model

IMAGE 5 **no outward** Exogen
CUS_EXP 3 no outward Endogen
PER_QUAL 7 no outward Endogen
PER_VAL 2 no outward Endogen
ECSI 3 no outward Endogen
CUS_COMP 1 no outward Endogen
CUS_LOY 3 no outward Endogen

*Mode A (LV-Mode = outward), 0 on program line 5,
1 latent variable per block (Deflate = no), 0 on program line 6*

24

=====

Etude Téléphonie mobile

Dimension No. 1

Partial Least-Squares Parameter Estimation

Change of Stop Criteria during Iteration

| Cycle No. | CR1 | CR2 | CR3 | CR4 | CR5 |
|-----------|-----------|------------|------------|------------|------------|
| 1 | .1026E+01 | .3231E+00 | .5287E+00 | .5147E+00 | .1773E-01 |
| 2 | .5873E-02 | .1176E-02 | .5439E-02 | .4155E-02 | -.3903E-03 |
| 3 | .9046E-04 | .5615E-04 | .1597E-04 | .2670E-04 | .2143E-04 |
| 4 | .2483E-05 | -.4172E-06 | -.1550E-05 | -.1371E-05 | .3166E-07 |

Convergence at Iteration Cycle No. 4

Path coefficients

```

=====
                IMAGE  CUS_EXP  PER_QUAL  PER_VAL  ECSI  CUS_COMP  CUS_LOY
-----
IMAGE           .000    .000    .000    .000    .000    .000    .000
CUS_EXP         .493    .000    .000    .000    .000    .000    .000
PER_QUAL        .000    .545    .000    .000    .000    .000    .000
PER_VAL         .000   -.066   -.540    .000    .000    .000    .000
ECSI            .153    .037    .544   -.200    .000    .000    .000
CUS_COMP        .000    .000    .000    .000    .540    .000    .000
CUS_LOY        -.212    .000    .000    .000   -.466   -.050    .000
=====
    
```

Correlations of latent variables

```

=====
                IMAGE  CUS_EXP  PER_QUAL  PER_VAL  ECSI  CUS_COMP  CUS_LOY
-----
IMAGE           1.000
CUS_EXP         .493    1.000
PER_QUAL        .731    .545    1.000
PER_VAL        -.508   -.360   -.576    1.000
ECSI            .671    .481    .791   -.604    1.000
CUS_COMP        .469    .250    .537   -.348    .540    1.000
CUS_LOY        -.548   -.366   -.524    .517   -.635   -.401    1.000
=====
    
```

Inner Model

```

=====
Block          Mean  Location  Mult.RSq  AvResVar  AvCommun  AvRedund
-----
IMAGE          5.2903  5.2903   .0000  207.3363   .4760   .0000
CUS_EXP        5.1199  2.5116   .2431  219.8322   .4711   .1145
PER_QUAL       5.2422  2.4514   .2971  149.0872   .5737   .1705
PER_VAL       -2.9955   .1743   .3351   73.8087   .8495   .2846
ECSI           4.6523   .2026   .6717   88.7433   .6825   .4585
CUS_COMP       2.6735   .1611   .2916    .0000   1.0000   .2916
CUS_LOY       -3.2648   .1574   .4318  433.2169   .5200   .2246
-----
Average                               .3244  185.5536   .5881   .1853
=====
    
```

Outer Model

```

=====
Variable      Weight      Loading  Location  ResidVar  Communal  Redundan
-----
  IMAGE      outward
IMAG1         .0145     13.5115    2.2992   172.8140    .5137    .0000
IMAG2         .0126     10.5838   19.3434   238.0597    .3200    .0000
IMAG3         .0136     15.5489  -18.4351   317.1061    .4326    .0000
IMAG4         .0176     16.1672  -12.3277   155.7097    .6267    .0000
IMAG5         .0144     12.0500   13.2760   152.9918    .4869    .0000
-----
  CUS_EXP    outward
CUEX1         .0231     12.3473    9.8955   170.9680    .4714    .1146
CUEX2         .0224     12.8096    6.9954   231.1035    .4152    .1009
CUEX3         .0253     16.9216  -15.2583   257.4248    .5266    .1280
-----
  PER_QUAL   outward
PERQ1         .0098     12.2630   12.8720    98.1148    .6052    .1798
PERQ2         .0085     13.6473   -2.7407   253.6826    .4234    .1258
PERQ3         .0118     16.1807  -10.3771   146.3491    .6414    .1906
PERQ4         .0094     13.9255    3.8455   141.5140    .5781    .1718
PERQ5         .0084     11.8000   14.4994   120.4663    .5361    .1593
PERQ6         .0095     13.8490    2.6912   134.8588    .5871    .1745
PERQ7         .0129     16.4124  -12.7914   148.6248    .6444    .1915
-----
  PER_VAL    outward
PERV1         -.0239   -22.5804  -10.3504    76.3317    .8698    .2914
PERV2         -.0247   -18.6053   10.0025    71.2857    .8292    .2779
-----
  ECSI       outward
CUSA1         .0158     9.7212   32.4203   92.6286    .5050    .3392
CUSA2         .0231    17.0633  -11.2935   92.0410    .7598    .5104
CUSA3         .0264    17.1403   -9.5627   81.5605    .7827    .5258
-----
  CUS_COMP   outward
CUSCO         .0397     25.2193    .0000     .0000     1.0000    .2916
-----
  CUS_LOY    outward
CUSL1         -.0185   -25.2005  -10.5839   234.0672    .7307    .3155
CUSL2         -.0061    -8.6089   16.2050   917.0478    .0748    .0323
CUSL3         -.0225   -21.3694    4.3239   148.5356    .7546    .3258
=====

```

Latent variables (Standardised)

```

=====
              IMAGE      CUS_EXP  PER_QUAL  PER_VAL      ECSI  CUS_COMP  CUS_LOY
-----
100293       -1.893      -.681    -1.633     2.180     -1.245    -.030     .715
100382        1.688      1.678     1.598    -1.869     1.292     1.292    -.905
100386       -1.016      -.400     -.630     -.247     -.123     -.470     .668
.
.
.
302284       -.245       .105     -.001     -.788     .170      .411     -.330
302291       -.092       -.650     -.782     .834     -.637    -2.233     .124
300589        1.008       .057     .927     -.266     .251      .851     -.973
=====

```