Institute of Food Research

OPLS: an ideal tool for interpreting PLS regression models?

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Contents

- Introduction
- Why use PLS instead of OPLS
 - Same predictive performance
 - PLS faster
 - Can use 1st PLS vector for 'OPLS interpretation'
- Look at the changes in X
- Conclusions

Introduction

- OPLS: PLS with integrated OSC filter



PLS regression – predict vector **y** from matrix **X**

- Model dimensionality defined by no. PLS factors, a
- Estimate a using cross
 validation
- Interpret regression model in terms of original variates

 $\mathbf{X}_{opls} = \mathbf{X} - \mathbf{T}_{o} \mathbf{P}_{o}^{\top} \qquad \mathbf{X} = \mathbf{T} \mathbf{P}^{\top} + \mathbf{E}_{pls} \qquad \mathbf{X} = \mathbf{T}_{p} \mathbf{P}_{p}^{\top} + \mathbf{T}_{o} \mathbf{P}_{o}^{\top} + \mathbf{E}_{opls}$

- Tweak 1: modify X_{opls}, e.g. rescale, derivatise
- Tweak 2: reduce amount filtered based on PCA of T_oP_o^T

Introduction

- OPLS: two properties

For y vector case and **no** tweaks
1. Predictions using *q* OPLS filter factors and *p* PLS regression factors are identical to predictions from a PLS model using *a* regression factors, where *a* = *q* + *p PLS and OPLS have same performance*

- 2. The 1st PLS latent vector is unaltered during OPLS 'filtering'.
 - 1st PLS vector can be interpreted as regression coefficients for OPLS filtered **X**

Main conclusion: build models using PLS instead of OPLS

Why use PLS instead of OPLS?

- Same predictive performance
- Avoid overfitting when a = 1
- PLS already standard chemometric tool
- PLS faster (e.g. SIMPLS)
- Simpler: 1 step versus 2 steps
- Can use 1st PLS vector for 'OPLS interpretation' ...
- Have choice of post-processing methods for further analysis

Why use PLS instead of OPLS? - Same predictive performance

OPLS filter factors PLS regression factors



- Have choice in partitioning model into filter and regression components
- Implications when interpreting X in terms of 'orthogonal', 'predictive', and 'residual' parts
- Convenient to use *a* 1 filter factors
- Alternative interpretation: model split into 'univariate weights' and 'multivariate advantage'

Why use PLS instead of OPLS? - Same predictive performance: who knew?



Literature survey of OPLS

133 papers citing original OPLS paper or 2 subsequent ones on O2PLS (inc. original)

99 surveyed in detail

70/133 by Umea or Imperial (53%)

55/99 in survey (56%)

Reasonably representative

• Bruwer et al 2007 Ind. Eng. Chem. Res. 46 864



Why use PLS instead of OPLS? - Same predictive performance: who knew? Good paper "The separation provided by OPLS-DA is particularly impressive and warrants further investigation in other proteomic studies." - Whelehan et al 2006 CILS 84 82 "Like PLS-DA, O-PLS-DA is a supervised pattern recognition technique, but Μ has improved predictive quality because the structured noise is modeled separately." - Want et al 2007 J Proteome Research 6 459 "The O-PLS-DA method provides a **prediction similar to that of PLS-DA**, but the interpretation of the models is improved because the structured noise is modeled separately from the variation common to the X and Y matrices." – Rezzi et al 2007 J Proteome Research 6 513 Tweaks can obscure equivalence "In OPLS, the group discrimination is forced to the first component, and thus classification results improved enormously as shown in Figure 4A and Possib 4B." - Wagner et al 2007 Anal. Chem 79 2918

Why use PLS instead of OPLS? - PLS faster

Can use SIMPLS rather than NIPALS • de Jong 1993 CILS **18** 251

Is speed important?

YES - for model validation

OPLS survey

34 OPLS-DA

9 PLS-DA with 'OPLS interpretation'

Need to evaluate lots of sub models:

- Estimation of performance (Use double-cross validation)
- Significance of summary stats. (Use y-scrambling)

Are PLS regression coefficients still important?

YES – to assess the model stability in the presence of ALL the systematic variability

- Convenient to use
- Relative weights variate subset independent
- Potential for updating
- 1st vector dependant on scaling
- Determines which variates look interesting





What to use?

Model using UV scaling – correlation based weights

- Plot covariance 'back scaling'
- 3. Colour code by correlation
- Look for variates which are both high in correlation and covariance
- Isn't this just a univariate based analysis?

"Differential metabogram"

- Martin et al 2007 J Proteome Research 6 1471



Holmes et al 2006 J Proteome Research 5 1313

11/14

What to use?

"Differential metabogram"

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Holmes et al 2006 J Proteome Research 5 1313

So far, build model using all the variates, then focus on interesting bits

Two PLS based suggestions

1. COVPROC

- Hoskuldsson 2001 CILS 55 23
- adds variates based on magnitude of 1st PLS vector
- can appraise subset model performance
- complements ranked lists of variates, eg GSEA

2. Powered PLS

Indahl 2005 J Chemom. 19 32

- Optimise the univariate weights
- Performs a restricted optimization of the weight vectors ... that passes through the PLS1 weight vector solution.
- Also has variate subset selection properties

Look at the changes in X — beer fermentation, variation in ABV (%)



Conclusions

- PLS is a tried and tested chemometric technique – don't ditch it just yet
- There is no performance advantage over PLS
- OPLS explicitly splits PLS model into multivariate advantage and univariate weight
- Can look at impact of filter on individual variates

 tangible representation of filter action
- OPLS better used for post processing rather than pre-filtering

And finally...

y = X βDirect approach $β = X^{-1} y$

Approximate \mathbf{X}^{-1} by \mathbf{X}^{T}

 $\mathbf{y}_{\text{new}} = \mathbf{x}_{\text{new-opls}} \times [\mathbf{X}^{\mathsf{T}}\mathbf{y}]$

OPLS compensates for approximating the inverse of X by its transpose

Least-squares method of normal equations

 $\beta = (\mathbf{X}^{\mathsf{T}}\mathbf{X})^{-1} \times \mathbf{X}^{\mathsf{T}}\mathbf{y}$ $\mathbf{y}_{\mathsf{new}} = \mathbf{x}_{\mathsf{new}} \times \begin{bmatrix} (\mathbf{X}^{\mathsf{T}}\mathbf{X})^{-1} & \mathbf{X}^{\mathsf{T}}\mathbf{y} \end{bmatrix}$ Multivariate advantage Univariate weight
OPLS acts as regularised inverse of the covariance matrix

CAC2008 Montpellier 30th June – 4th July 2008