27411 Chemometrics and Biological Data Analysis

Lecture 6: PLS - Partial Least Squares (Regression)

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What is PLS?

- PLS1: Many X-variables, One Y-variable
 - Alternative to PCR
 - Uses Y-information to construct X-components
- 2 PLS2: Many X-variables, Many Y-variables
 - ullet Relates components of X to components of Y

Overview

What is PLS?

2 PLS - How to do it! (same as for PCR)

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PLS1 method:

- Centering and (possibly) scaling.
- **2** Do p simple regressions: x_1 on y, x_2 on y, ..., x_p on y.
- Use these p regression coefficients (w_1, w_2, \ldots, w_n) to define the first PLS-component.

$$\boldsymbol{t}_1 = w_1 \boldsymbol{x}_1 + w_2 \boldsymbol{x}_2 + \cdots w_p \boldsymbol{x}_p$$

- Do the simple regression of y on t_1
- ullet Do the simple regressions of x_1, x_2, \ldots, x_n on t_1
- Repeat 2-5 on residuals for as many components as wanted/needed.
- Choose number of components by Cross-Validation

PLS1

- PLS is a canonical covariance method
- PLS1 finds X-components with maximal Y-covariance:

$$\max_{||lpha||=1} \mathsf{Cov}^2(oldsymbol{y}, oldsymbol{X}lpha)$$

• Or equivalently:

$$\max_{||\alpha||=1} \mathsf{Corr}^2(\boldsymbol{y}, \boldsymbol{X}\alpha) \mathsf{Var}(\boldsymbol{X}\alpha)$$

- PCR is a canonical variance method
- PCR finds X-components with maximal Y-variance:

$$\max_{||lpha||=1} \mathsf{Var}(oldsymbol{X}lpha)$$

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How to do it?(same as for PCR)

- Explore data
- Do modelling (choose number of components, consider variable selection)
- Validate (residuals, outliers, influence etc)
- 1 Iterate e.g. on 2. and 3.
- Interpret, conclude, report.
- If relevant: predict future values.

PLS versus PCR

- PLS uses the y-information for building components, PCR does not
- PLS and PCR often predicts on a similar level of error
- PLS often does so with fewer components

Cross Validation ("Full")

- Leave out one of the observations
- Fit a model on the remaining(reduced) data
- Predict the left out observation by the model: $\hat{y}_{i,val}$
- Do this in turn for ALL observations AND calculate the overall performance of the model:

$$\mathsf{RMSEP} = \sqrt{\sum_{i}^{n} (y_i - \hat{y}_{i,val})^2 / n}$$

(Root Mean Squared Error of Prediction)

Cross Validation ("Full")

Choose the optimal number of components:

- The one with overall minimal error
- The first local mininum
- In Hastie et al: the smallest number within the uncertainties of the overall minimum one.

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PLS - How to do it! (same as for PCR)

Cross Validation - principle

• Minimizes the expected prediction error:

Squared Prediction error = $Bias^2 + Variance$

- Including "many" PLS-components: LOW bias, but HIGH variance
- Including "few" PLS-components: HIGH bias, but LOW variance
- Choose the best compromise!
- Note: Including ALL components = MLR (when n > p)

Resampling

- Cross-Validation (CV)
- Jackknifing (Leave-on-out CV)
- Bootstrapping
- A good generic approach:
 - Split the data into a TRAINING and a TEST set.
 - Use Cross-validation on the TRAINING data
 - Check the model performance on the TEST-set
 - MAYBE: REPEAT all this many times (Repeated Double Cross Validation)

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Validation - exist on different levels

- Split in 3: Training(50%), Validation(25%) and Test(25%)
 - Requires many observations Rarely used
- Split in 2: Calibration/training (67%) and Test(33%) us CV/bootstrap within the training
 - more commonly used
- No "fixed split", but repeated splits by CV/bootstrap, and then CV within each training set ("Repeated double CV")
- No split, but using (one level of) CV/bootstrap.
- Just fitting on all and checking the error.

Overview

- What is PLS?
- 2 PLS How to do it! (same as for PCR)

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