

HANDLING RANK DEFICIENCY IN UV/VIS SPECTRA FROM KINETIC STUDIES BY VARIOUS MCR TECHNIQUES

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Background

Monitoring UV/vis absorption is one of the oldest instruments for studying chemical reactions. It combines versatility, high sensitivity, low response time etc. *Multivariate curve resolution* (MCR) is usually employed to resolve highly overlapping absorption spectra of the components:

Given X is the data $M \times N$ matrix

$$X = C_0 S_0 + E,$$

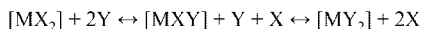
C_0 is the concentrations ($M \times A$) matrix, S_0 is the spectra ($N \times A$) matrix, E is the noise matrix. Matrices C_0 , S_0 are subject to *chemical constrains*.

MCR aims to find A (number of substances), and C_X and S_X close to C_0 , S_0 in some sense. Usua. a least-square (LS) sense is used:

$$\|X - C_X S_X\|_2 = \min$$

Problem

Many important reactions in coordination chemistry have a following generic mechanism:



M here is coordination center, X and Y are some ligands.

Most intensive UV/vis bands usually originate from M-X and M-Y bonds. So, we quite often observe:

$$S_{[MXY]} \approx \frac{1}{2}(S_{[MX_2]} + S_{[MY_2]})$$

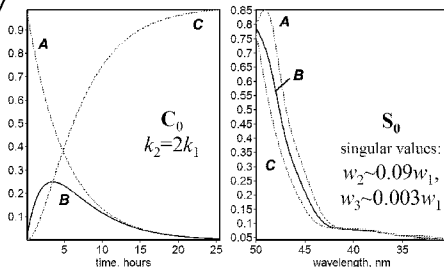
Hence, the efficient rank of matrices S_0 and X is less, than the number of independent concentrations => we observe additional *rank deficiency*.

This type rank deficiency is not obvious *a priori*, but its is not as practically rare as it may seem.

AIM: Crash-test various MCR approaches against "ill-conditioned" S_X matrices.

Model Data

Model reaction: $A \rightarrow B, k_1; B \rightarrow C, k_2$



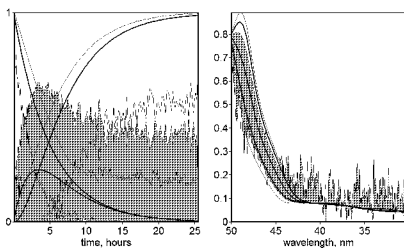
k_1 and S_0 were taken to directly emulate 2-step hydrolysis reaction of anticancer drug cisplatin. $M = 175, N = 247$.

SOFT only

Direct *Alternating least squares* approach (MCR-ALS)



1. LS-estimate spectra $\hat{S} = X^T C (C^T C)^{-1}$
2. Correct \hat{S} for nonnegativity, $\hat{S} \rightarrow S$
3. LS-estimate concentrations $\hat{C} = X S (S^T S)^{-1}$
4. Correct \hat{C} for closure, windows and nonneg., $\hat{C} \rightarrow C$
5. Loop (1) until convergence on $\|X - CS\|_2$



98% converged to $\epsilon 10^{-4}$, $\langle L.S.Eval \rangle = 530$

SOFT-outside & HARD-inside

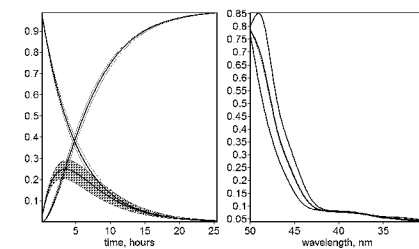
Alternating least squares constrained with *kinetic model*



1. LS-estimate spectra $\hat{S} = X^T C (C^T C)^{-1}$
2. Correct \hat{S} for nonnegativity, $\hat{S} \rightarrow S$
3. LS-estimate concentrations $\hat{C} = X S (S^T S)^{-1}$
 - 4a. Generate C from *kinetic model*
 - 4b. NLSF: loop (4a) until convergence on $\|\hat{C} - C\|_2$
5. Loop (1) until convergence on $\|X - CS\|_2$

Minimum/Maximum Bands of the Solution, $k_2 = 2k_1$

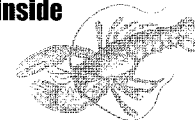
100 replicas, noise 0.3%. Band for B is painted gray



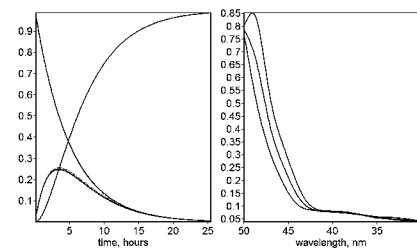
100% convgd. to $\epsilon 10^{-4}$, $\langle L.S.Eval \rangle = 157$, $\langle Kin.Eval \rangle = 613$

HARD-outside & SOFT-inside

Full-matrix non-linear least squares fit



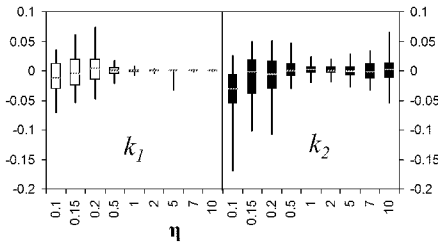
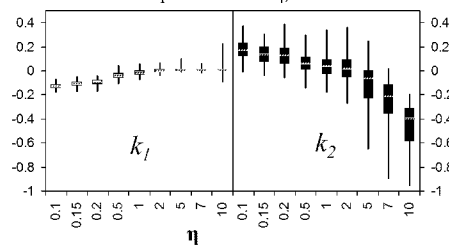
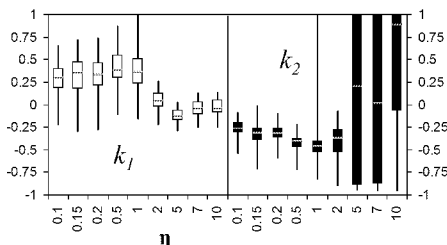
1. Generate C from *kinetic model*
2. LS-estimate spectra $\hat{S} = X^T C (C^T C)^{-1}$
3. Correct \hat{S} for nonnegativity, $\hat{S} \rightarrow S$
4. NLSF loop (1) until convergence on $\|X - CS\|_2$



100% convgd. to $\epsilon 10^{-4}$, $\langle L.S.Eval \rangle = 18$, $\langle Kin.Eval \rangle = 19$

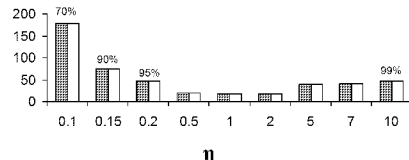
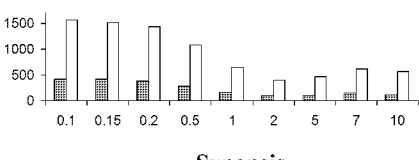
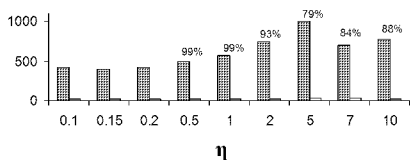
Quartile Plots of Relative Errors for k_1 and k_2 Prediction, $k_2 = (1+\eta)k_1$

100 replicas for each η , noise 0.3%



Convergence Expenses, Gray Bars – <Full-matrix LS-fits>, White Bars – <Kinetic Model Evaluations>

% of converged cases is indicated if less than 100%



WRECKED

Synopsis
BARELY PASSABLE

TOLERABLE