

Multi-way Analysis of 2D Liquid Chromatographic Metabolomics Data

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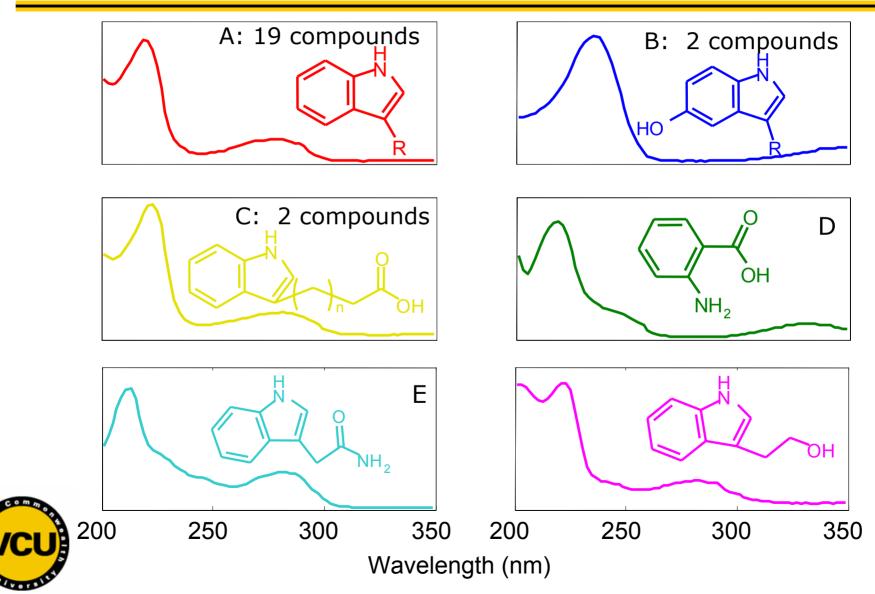
Jerry D. Cohen University of Minnesota, Department of Horticultural Science Indole-3-Acetic Acid (IAA) in Plants

- Primary growth hormone responsible for cell division and elongation, flowering, root initiation, fruit ripening, promoting vascular tissue growth, controlling premature abscission of leaves and fruit
- Synthesized from tryptophan by a variety of pathways
- Metabolic Profiling the identification and quantification of a selected group of metabolites in a biological system
- Metabolomics unsupervised comparison of different biological samples to elucidate differences in metabolite levels
- 2DLC with diode array detection (DAD) was used to compare mutant and wild type maize samples to 26 indolic metabolite standards

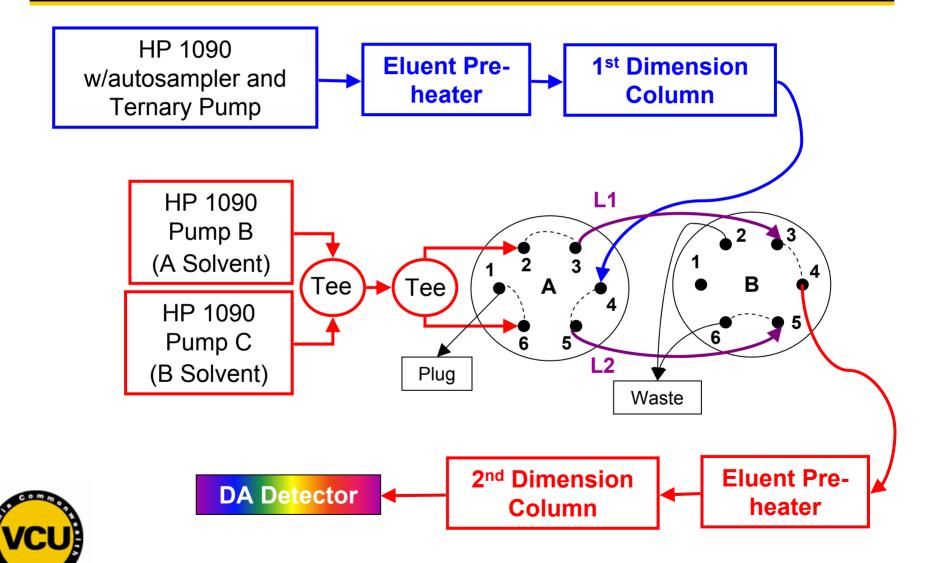


Wright, et al. *Science*, **1991**, *254*, 998-1000 Fiehn, O. *Plant Mol. Biol.* **2002**, *48*, 155-71

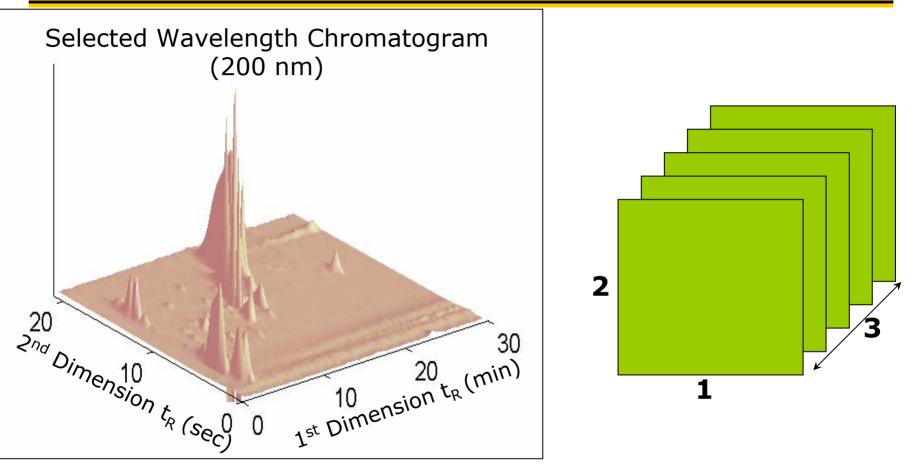
6 spectral components of 26 indolic standards



2DLC Instrumentation Capable of Gradient Elution in Both Dimensions



2DLC Data Structure – Three Way Data

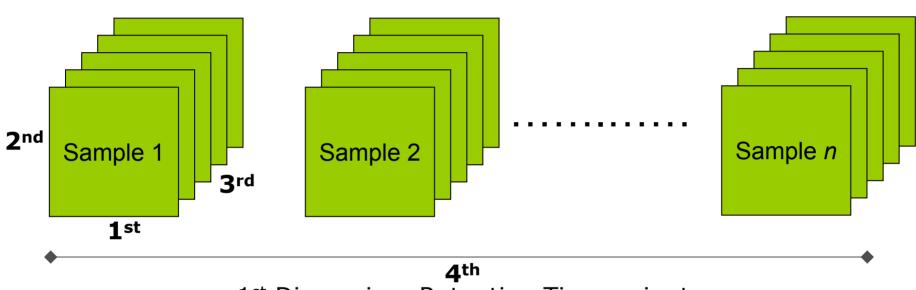




Data Dimensions:

- 1. 1st Dimension Retention Time (min)
- 2. 2nd Dimension Retention Time (sec)
- 3. Wavelength (nm)

Four-Way Quadrilinear Data



1st Dimension, Retention Time, minutes
2nd Dimension, Retention Time, seconds
3rd Dimension, Wavelength, nm
4th Dimension, Sample number



The instrument response of a pure component in all domains is unique, consistent, and independent of the presence of other species

Booksh, K. S. et al., Anal. Chem. 1994, 66, 2561-69.

Description of Samples

- Mobile phase blank
- Standard mixture containing 26 indoles
- Duplicate wild type maize seedling samples
- Duplicate *orp* mutant maize seedling samples
 - Lacks gene for tryptophan synthase β
 - IAA is produced via tryptophan-independent pathway



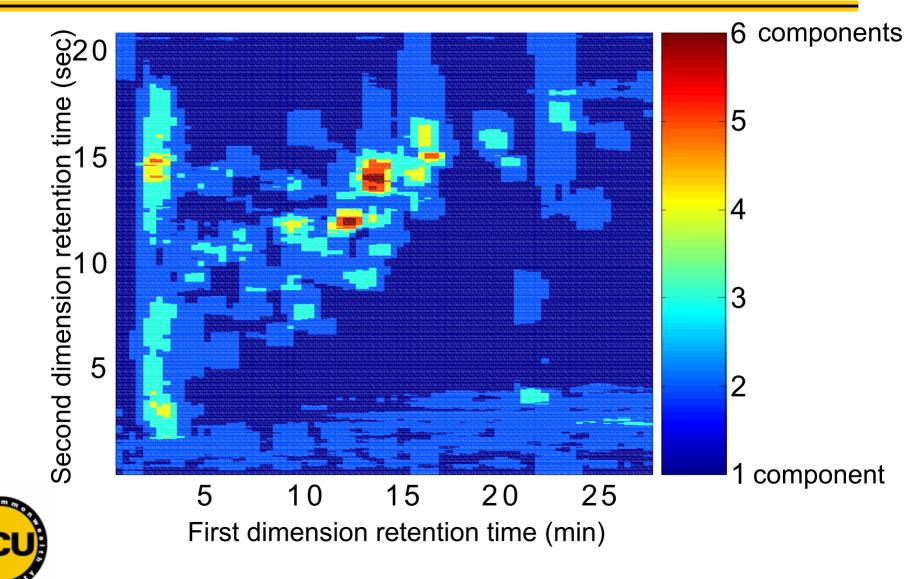
Fixed Size Image Window – Evolving Factor Analysis^a

- FSIW-EFA uses sections of an image and performs factor analysis on a moving window
- Rank information is *local* results estimate the complexity of an image for exploratory analysis
- Traditional EFA approaches would require unfolding of the three-way data set and loss of complex spatial structure

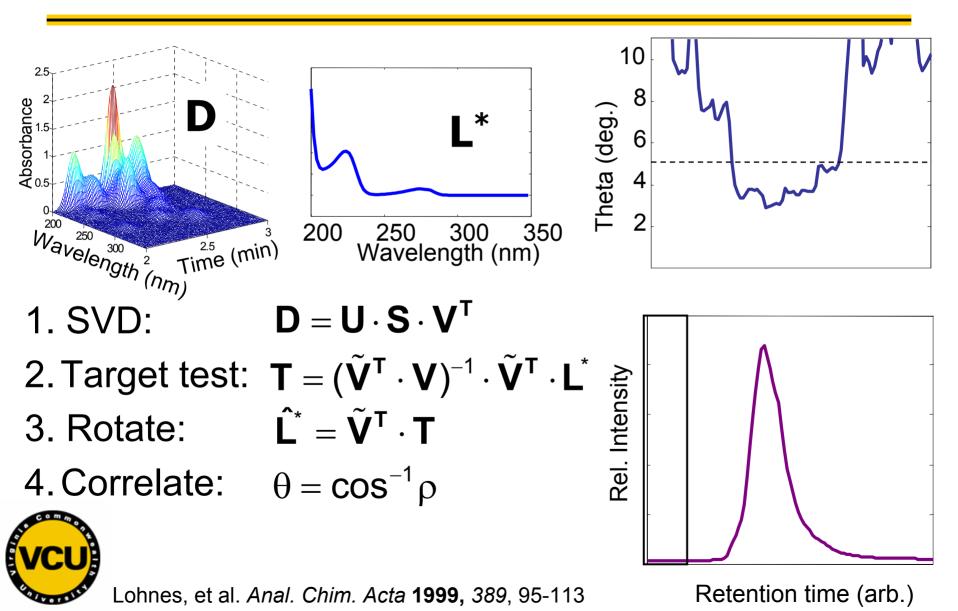
 Can be used to select sections of data for subsequent analysis

^ade Juan, A. et al. Chemom. Intell. Lab. Syst., 2005, 77, 2005, 64-74

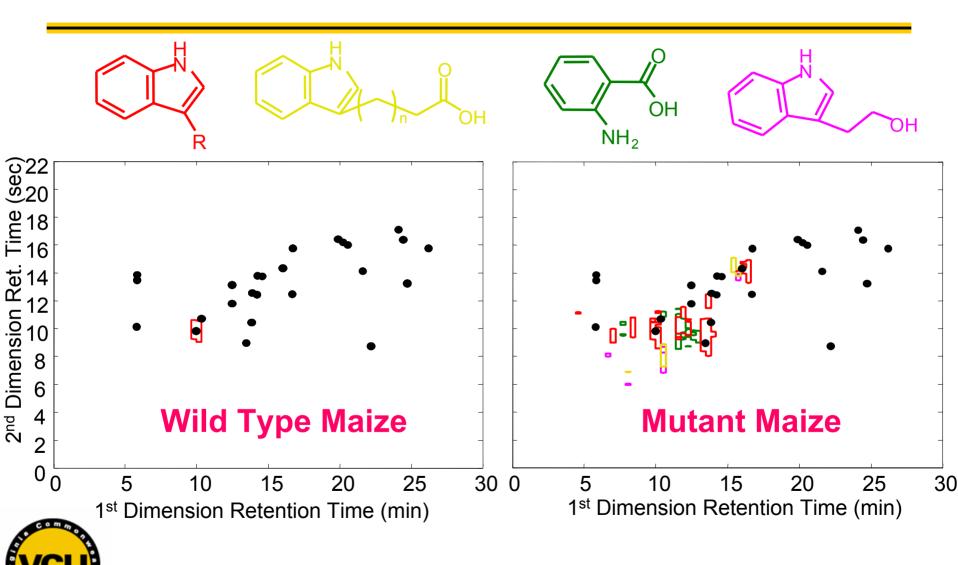
Summed Rankmap: Standards, Wild Type, Mutant



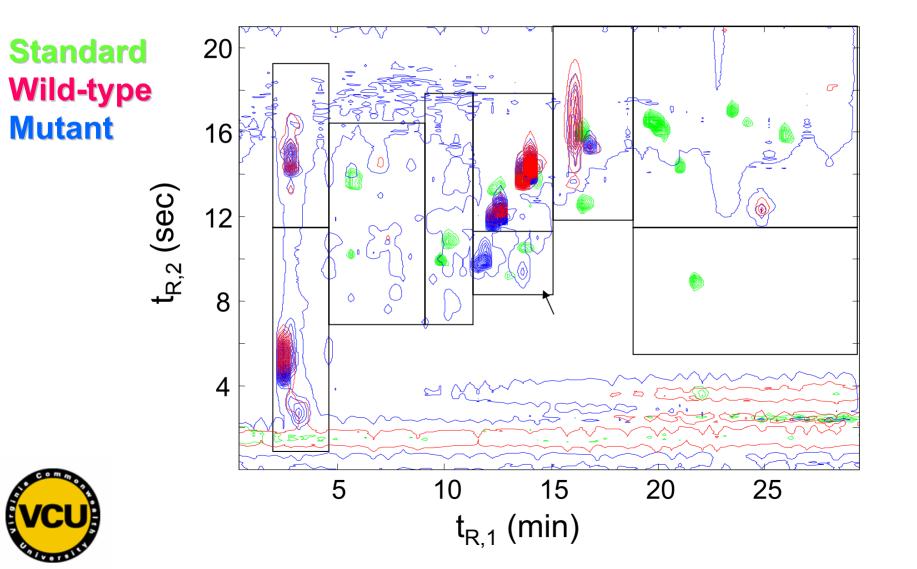
Window Target Testing Factor Analysis



Results of Qualitative Analysis



2D-Chromatograms – 220 nm



Modeling with PARAFAC and fALS

The **four-way** PARAFAC model is represented mathematically as a sum over all of the elements of each mode (where *c* is the rank of the data)^a: $d_{hijk} = \sum_{f=1}^{c} W_{hf} X_{if} Y_{jf} Z_{kf} + e_{hijk}$

The PARAFAC model is solved using alternating least squares (ALS)

ALS with flexible constraints (fALS)^b allows the selective application of the unimodality constraint to selected components



^aAndersson, C. A.; Bro, R. *Chemom. Intell. Lab. Syst.* **2000,** *52*, 1-4 ^bBezemer, E.; Rutan, S. C. *Chemom. Intell. Lab. Syst.* **2006,** *81*, 82-93

 $D = X \cdot Y^{T}$ $X = (Y^{T})^{\dagger} \cdot D$ $Y^{T} = D \cdot (X)^{\dagger}$

PARAFAC-ALS vs. fALS

- PARAFAC-ALS^a
 - Multilinearity required
 - Constraints provide LS solutions with guaranteed convergence
 - Constraints must be applied to all components

• fALS^b

- Multilinearity optional
- Constraints may be adhoc, but not LS optimal
- Constraints can be applied to selected components



^aAndersson, C. A.; Bro, R. *Chemom. Intell. Lab. Syst.* **2000,** *52*, 1-4 ^bBezemer, E.; Rutan, S. C. *Chemom. Intell. Lab. Syst.* **2006,** *81*, 82-93

Data Analysis Procedure

Section data according to local complexity

Rank determination in each section

PARAFAC-ALS^a – SVD initiated, non-negativity

fALS^b – initiated with previous, non-negativity

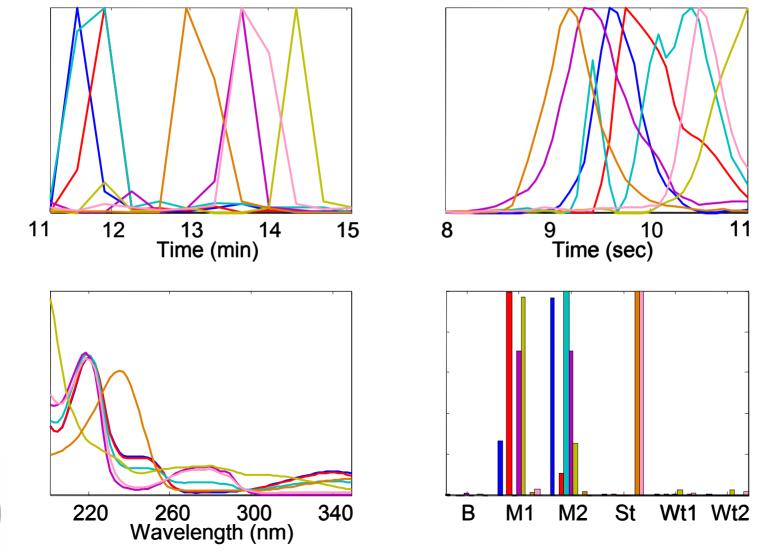
fALS^b – initiated with previous, unimodality on nonbackground components

PARAFAC – ALS^a – initiated with previous, non-negativity



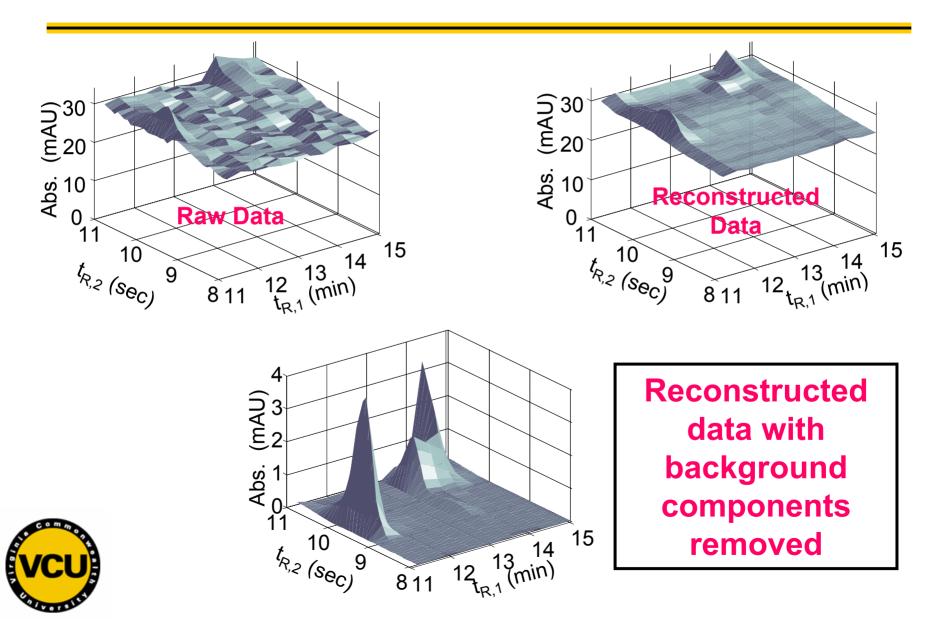
^aAndersson, C. A.; Bro, R. *Chemom. Intell. Lab. Syst.* **2000,** *52*, 1-4 ^bBezemer, E.; Rutan, S. C. *Chemom. Intell. Lab. Syst.* **2006,** *81*, 82-93

PARAFAC Results for Selected Section

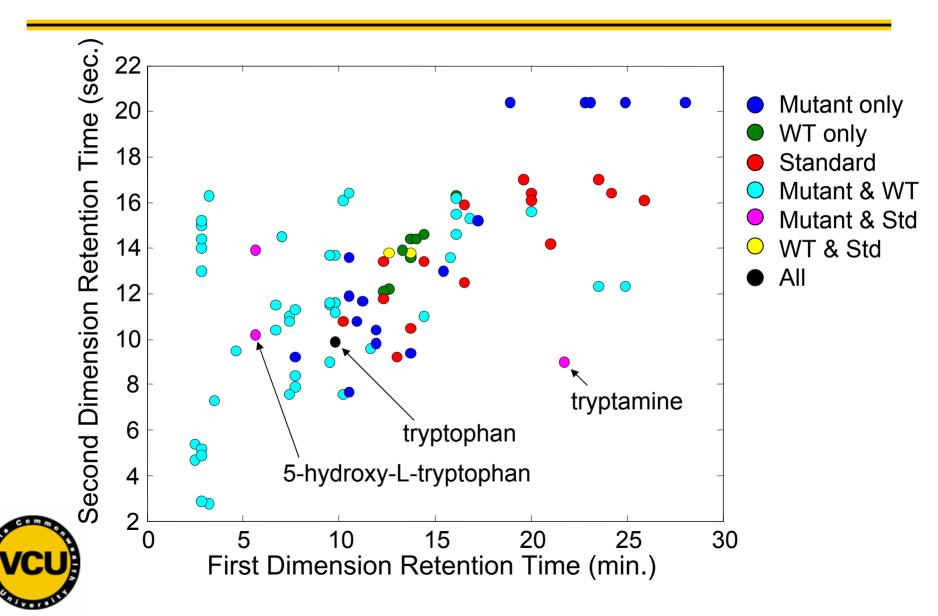




Reconstructed PARAFAC Results – Wt1



Results of Analysis of Entire Data Set



Quantitative Results of PARAFAC Analysis

- 95 distinct chromatographic peaks were resolved
- Many peaks showed differential expression between wild type and mutant samples

	Mutant 1	Mutant 2	WT1	WT2	
5-hydroxy-L- tryptophan	ND*	0.6	ND*	ND*	
Tryptophan	1.4	1.1	1.8	1.6	
Indole-3-acetyl- L-alanine	ND*	ND*	0.8	0.3	
Tryptamine <u>*ND – not detect</u>	1.9 ed	ND*	ND*	ND*	
Quantitative res	Quantitative results are in μg of indole per gram of plant material				

Selectivity in Multi-way Analysis^a

Messick, Kalivas, Lang (MKL)^b
 ▶PARAFAC

➢ Appropriate when all components are calibrated
 ➢ SEL_n = {[(X^TX)*(Y^TY)]⁻¹}_{nn}^{-1/2}

Ho, Christian, Davidson (HCD)^c
 ➢GRAM

> Appropriate when only the target analyte is calibrated > SEL_n = {[(X^TX)⁻¹]_{nn}[(Y^TY)⁻¹]_{nn}}^{-1/2}

^aOlivieri, A. C., *Anal. Chem.* **2005**, 77, 4936-3946 ^bMessick, N. J.; Kalivas, J. H.; Lang, P. M. *Anal. Chem.* **1996**, 68, 1572-1579 ^cHo, C.-N.; Christian, G. D.; Davidson, E. R., *Anal. Chem.* **1980**, 52, 1071-1079

Selectivity in Multi-way Analysis

- The selectivity calculations predict the relative decrease in precision that is observed relative to that observed for a pure sample.
- Olivieri observed that for some multi-way situations, neither selectivity formulation predicted the results of Monte Carlo calculations.
 - SEL_{MKL} upper limit of selectivity
 - SEL_{HCD} lower limit of selectivity



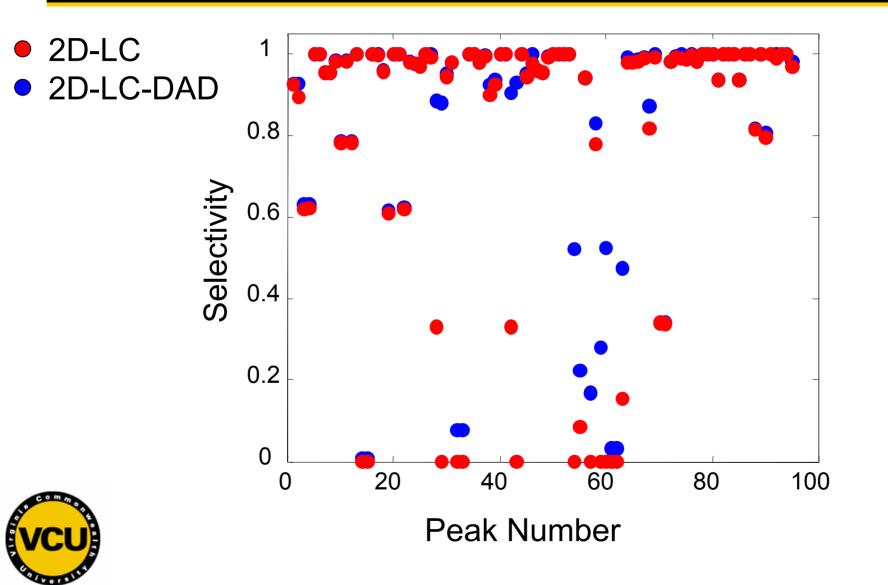
Olivieri, A. C., Anal. Chem. 2005, 77, 4936-3946

Calculation of Selectivity for 2DLC vs. 2D-LC-DAD

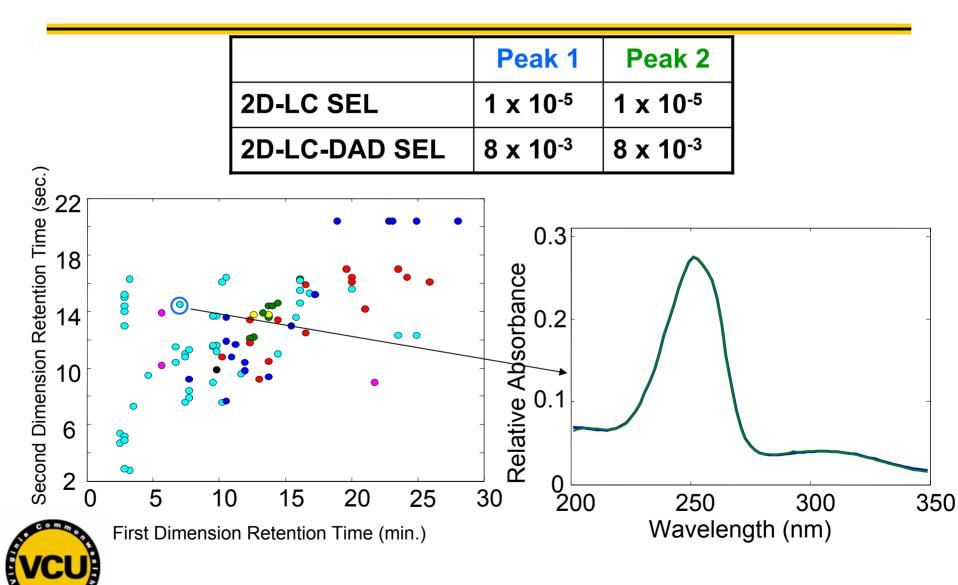
- X, Y, and Z simulated to approximate the real data
- X and Y 1st and 2nd dimension retention profiles, using resolved retention times and simulated, Gaussian peaks
- **Z** spectral profiles as resolved by PARAFAC
- Background components omitted from the analysis



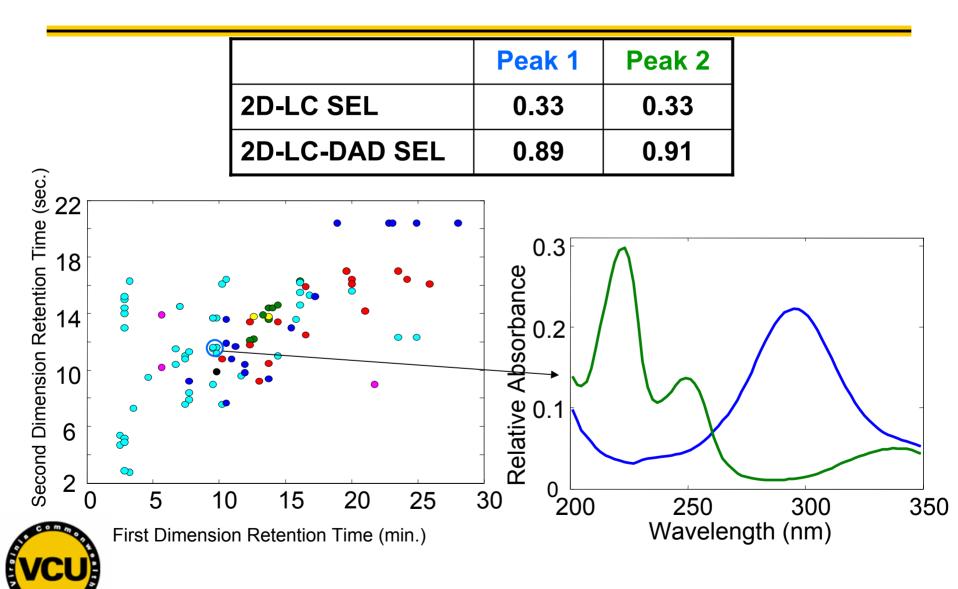
MKL Selectivity



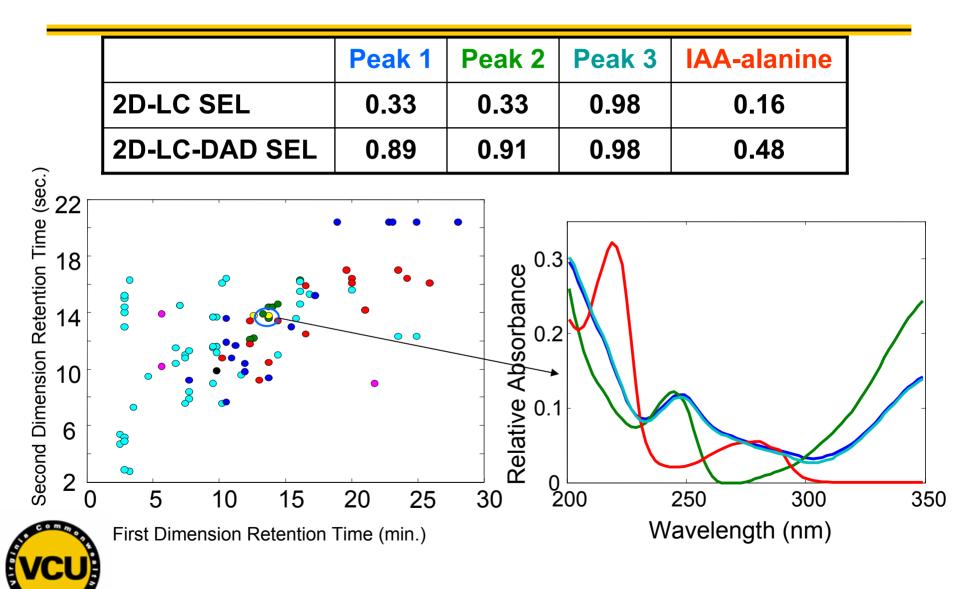
MKL Selectivity – 2D-LC vs. 2D-LC-DAD



MKL Selectivity – 2D-LC vs. 2D-LC-DAD



MKL Selectivity – 2D-LC vs. 2D-LC-DAD



MKL Selectivity Comparisons

Data Dimensions	Average selectivity per component	Peak Capacity	
Column 1, single wavelength	0.15	50	
Column 1 + DAD	0.36	n.a.	
Column 2, single wavelength	0.05	17.4	
Column 2 + DAD	0.25	n.a.	
2D-LC, single wavelength	0.78	870	
2D-LC-DAD	0.84	n.a.	



Conclusions

- 1. Three chemometric methods (WTTFA, PARAFAC-ALS, and fALS) have been applied to four-way quadrilinear data generated by running multiple samples with 2D-LC-DAD.
- 2. These methods result in a great enhancement in S/N and background suppression.
- 3. Several indole conjugates have been identified in mutant and wild type maize samples.
- 4. The indole content of the wild type and mutants are clearly differentiated.
- 5. The quantitative capabilities of multi-way modeling have been demonstrated.
- 6. Multivariate selectivity has been shown to relate to chromatographic figures of merit.



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