



# NMR Data Pre-processing

## UAB Metabolomics Training Course

July 17-21, 2016

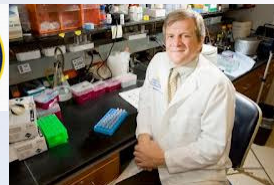
Wimal Pathmasiri, Rodney Snyder  
NIH Eastern Regional Comprehensive Metabolomics Resource Core  
(RTI RCMRC)

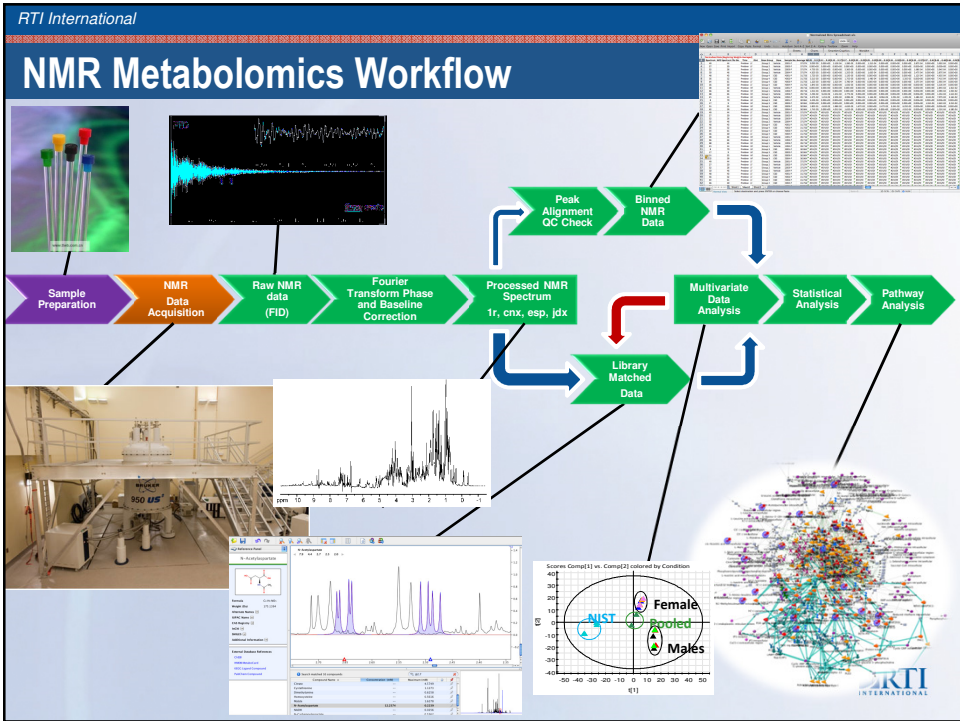
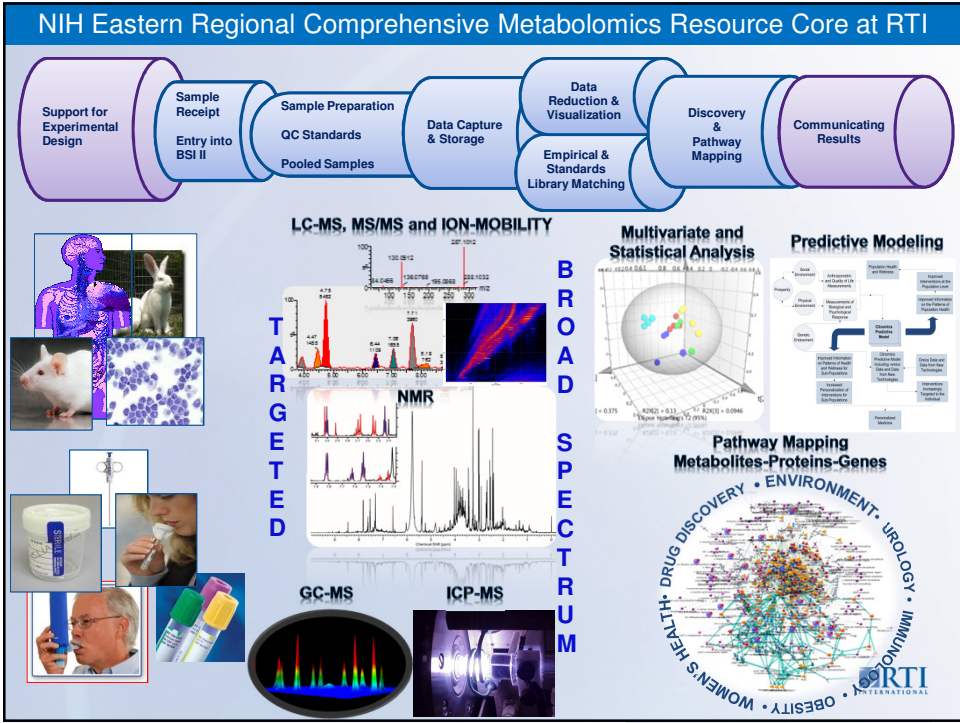
RTI International is a trade name of Research Triangle Institute.

[www.rti.org](http://www.rti.org)

## NIH Common Fund Metabolomics Cores

NIH Metabolomics Centers Ramp Up | November 4, 2013 Issue - Vol. 91 Issue 44 | Chemical & Engineering News. by Jyllian Kemsley





## Data Pre-processing

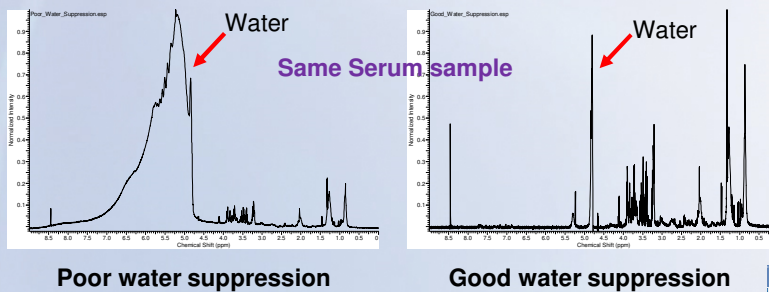
- After NMR data acquisition, the result is a set of spectra for all samples.
- For each spectrum, quality of the spectra should be assessed.
  - Line shape
  - Phase
  - Baseline
- Spectra should be referenced
  - Compounds commonly used: DSS, TSP, Formate
- Variations of pH, ionic strength of samples has effects on chemical shift
  - Peak alignment
  - Bucket integration
- Remove unwanted regions

## Quality Control Steps

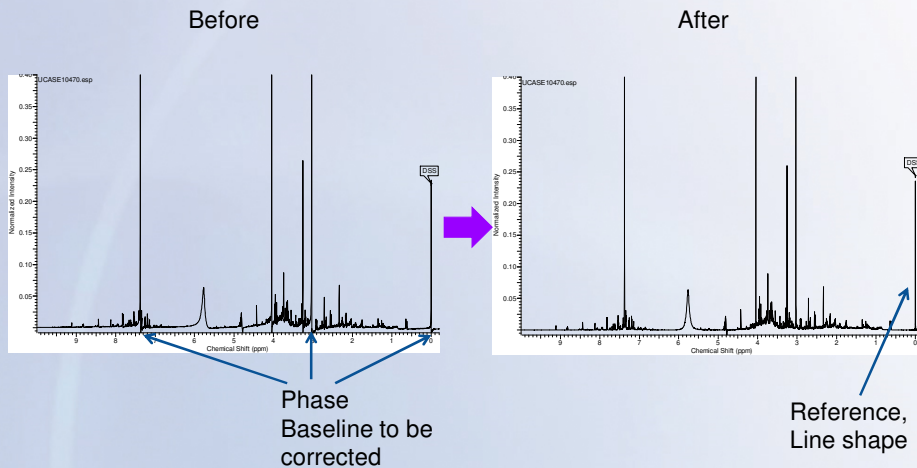
- Quality of metabolomics analysis depends on data quality
- Typical problems
  - Water peak (suppression issues)
  - Baseline (not set at zero and not a flat line)
  - Alignment of peaks (chemical shift, due to pH variation)
  - Variation in concentration (eg. Urine)
- High quality of data is needed for best results

## Water Suppression Effects and Other Artifacts

- If water is not correctly suppressed or removed there will be effects on normalization
- Need to remove other artifacts
- Remove drug or drug metabolites



## NMR Pre-processing



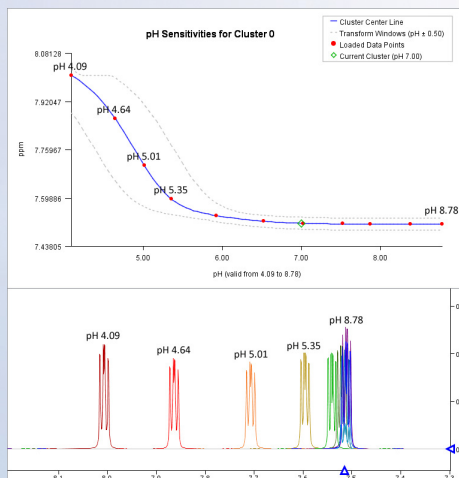
## pH Dependence of Chemical Shift

## Chemical shift variability

- pH
- ionic strength
- metal concentration

## Methods to overcome this problem

- Use a buffer when preparing samples
- Binning (Bucketing)
  - Fixed binning
  - Intelligent binning
  - Optimized binning
- Available data alignment tools
  - Recursive Segment-wise Peak Alignment (RSPA)
  - Icoshift
  - speaq



<http://www.chenomx.com/software/software.php>

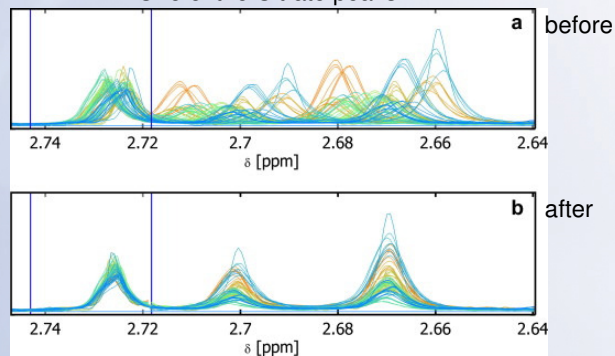
Savorani, F. et al., Journal of Magnetic Resonance, Volume 202, Issue 2, 2010, 190 – 202  
 Vu, T. N. et al., BMC Bioinformatics 2011, 12:405

## Peak Alignment

## Example

## icoshift

## One of the Citrate peaks

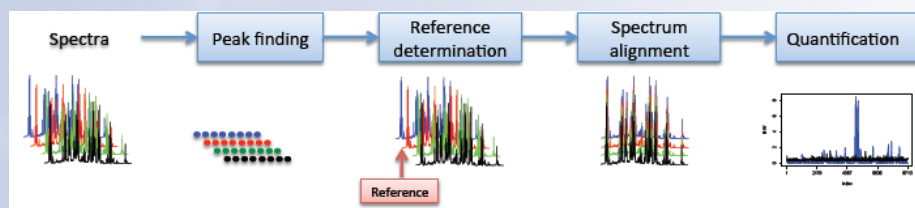


Savorani, F. et al., Journal of Magnetic Resonance, Volume 202, Issue 2, 2010, 190 - 202

## Peak Alignment

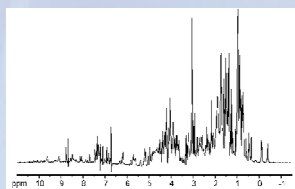
Example

speaq

Vu, T. N. et al., *BMC Bioinformatics* 2011, 12:405

## NMR Binning

- A form of quantification that consists of segmenting a spectrum into small areas (bins/buckets) and attaining an integral value for that segment
- Binning attempts to minimize effects from variations in peak positions caused by pH, ionic strength, and other factors.
- Two main types of binning
  - Fixed binning
  - Flexible binning

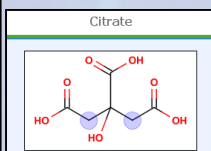


A screenshot of a software window titled 'Normalized Bin Statistics.xls'. It displays a large table with columns for 'Bin', 'Area', 'Height', 'Width', 'Mean', 'StdDev', 'Min', 'Max', 'Integral', and 'Integral/Height'. The table contains numerous rows of numerical data representing the results of NMR binning.

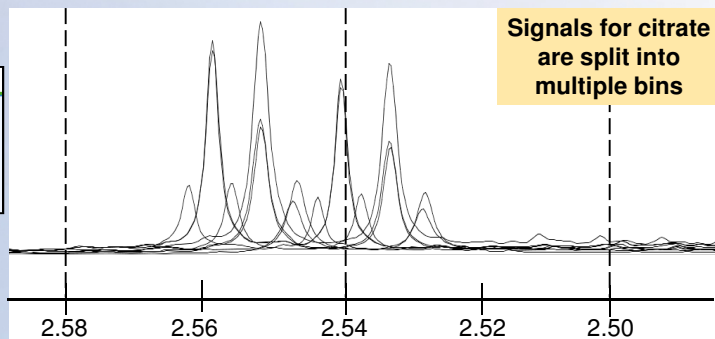
## NMR Binning

Peak shift can cause the same peak across multiple samples to fall into different bins

- The entire NMR spectrum is split into evenly spaced integral regions with a spectral window of typically 0.04 ppm.
- The major drawback of fixed binning is the non-flexibility of the boundaries.
- If a peak crosses the border between two bins it can significantly influence your data analysis

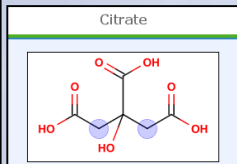


Fixed Binning

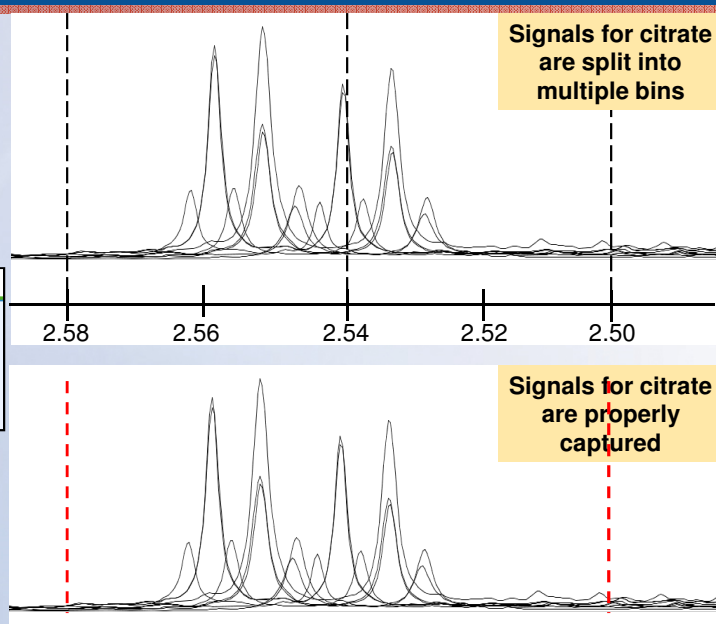


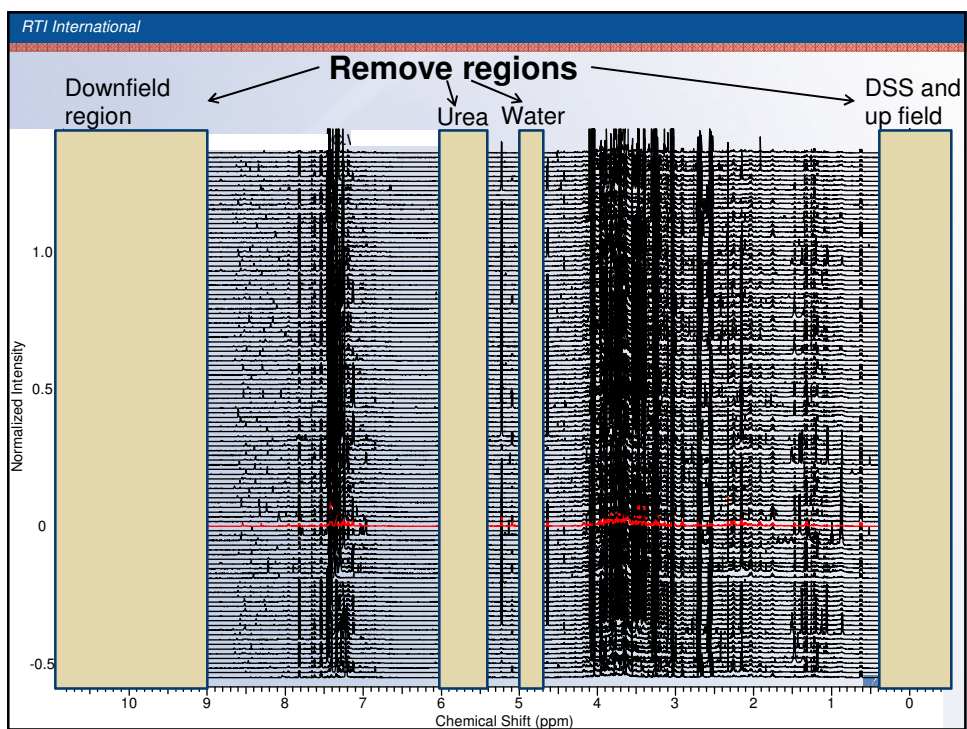
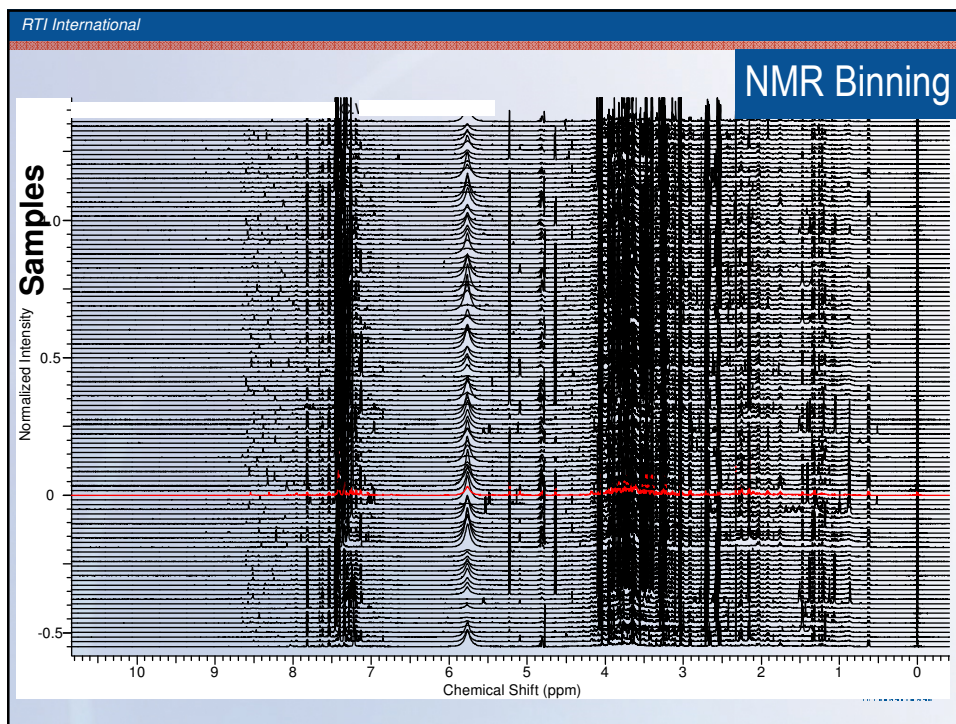
## NMR Binning

Fixed Binning



Smart Binning







## Binning

- Integrate bins (0.04 ppm bin size)
- Normalize integral of each bin to the total integral of each spectrum
- Merge metadata
- Result is a spreadsheet ready for further multivariate data analysis and other statistical analysis

Sample ID	Disease Group	[0.40 .. 0.46]	[0.46 .. 0.52]	[0.52 .. 0.54]	[0.54 .. 0.57]	[0.57 .. 0.60]	[0.60 .. 0.66]	[0.66 .. 0.68]	[0.68 .. 0.71]	[0.71 .. 0.75]
C0559	Cases	7.60E-05	0.00E+00	7.32E-02	8.48E-02	3.20E-02	1.84E+00	1.31E-01	3.60E-01	3.67E-01
C0629	Cases	0.00E+00	1.78E-02	0.00E+00	2.18E-02	0.00E+00	1.08E+01	0.00E+00	0.00E+00	3.02E-02
C0640	Cases	3.44E-04	0.00E+00	1.83E-03	1.86E-04	0.00E+00	4.51E+00	0.00E+00	0.00E+00	0.00E+00
C0835	Cases	6.41E-04	0.00E+00	6.44E-03	0.00E+00	3.96E-03	3.28E+00	0.00E+00	5.12E-03	1.75E-02
D0613	Cases	6.63E-03	0.00E+00	0.00E+00	1.06E-02	0.00E+00	5.79E+00	0.00E+00	6.36E-02	3.02E-01
D0762	Cases	0.00E+00	0.00E+00	1.79E-02	1.98E-02	0.00E+00	9.37E+00	0.00E+00	0.00E+00	1.74E-02
D1113	Cases	3.14E-03	2.42E-03	8.02E-02	1.04E-01	5.32E-03	3.74E+00	0.00E+00	2.02E-02	1.84E-01
D1158	Cases	0.00E+00	3.71E-03	2.35E-02	4.83E-02	0.00E+00	5.02E+00	0.00E+00	1.91E-02	0.00E+00
D2090	Cases	0.00E+00	0.00E+00	2.45E-03	9.98E-04	0.00E+00	5.76E+00	0.00E+00	1.24E-02	1.04E-02
E0004	Cases	1.72E-03	0.00E+00	6.85E-02	3.05E-02	0.00E+00	1.47E+00	6.90E-02	3.61E-01	4.08E-01
E0195	Cases	0.00E+00	1.69E-03	5.57E-02	6.29E-02	0.00E+00	2.77E+00	1.34E-01	2.04E-01	4.56E-01
E0252	Cases	1.25E-03	0.00E+00	1.05E-03	1.03E-02	0.00E+00	9.17E+00	0.00E+00	1.08E-02	2.30E-02
E0609	Cases	4.11E-03	0.00E+00	2.23E-02	1.34E-03	3.18E-03	3.54E+00	0.00E+00	3.28E-02	9.09E-01
E0487	Cases	1.72E-03	0.00E+00	0.00E+00	1.00E-02	0.00E+00	4.00E+00	0.00E+00	1.36E-02	0.00E+00
F0036	Cases	1.66E-02	0.00E+00	0.00E+00	2.06E-02	0.00E+00	1.22E+01	1.04E-02	0.00E+00	5.97E-01
F0108	Cases	0.00E+00	2.31E-03	6.30E-03	1.11E-02	0.00E+00	7.17E+00	0.00E+00	1.65E-02	2.21E-01
A0233	Control	0.00E+00	1.86E-02	0.00E+00	1.82E-02	0.00E+00	1.61E+01	0.00E+00	2.91E-03	0.00E+00
A0490	Control	0.00E+00	0.00E+00	2.99E-03	3.60E-02	0.00E+00	2.97E+00	0.00E+00	4.00E-02	5.46E-01
A2003	Control	0.00E+00	0.00E+00	3.45E-02	2.20E-02	0.00E+00	1.80E+00	0.00E+00	0.00E+00	0.00E+00
C0586	Control	0.00E+00	1.69E-02	0.00E+00	6.64E-03	0.00E+00	1.92E+01	0.00E+00	6.51E-02	0.00E+00
C2177	Control	0.00E+00	0.00E+00	3.02E-02	3.59E-02	0.00E+00	2.35E+00	0.00E+00	3.19E-02	1.49E-01
D0177	Control	9.21E-03	0.00E+00	1.69E-02	1.47E-02	0.00E+00	2.43E+00	0.00E+00	4.46E-02	0.00E+00
D0729	Control	0.00E+00	1.88E-03	5.58E-02	7.87E-02	2.92E-02	3.16E+00	6.59E-02	2.80E-01	4.30E-01
D0909	Control	0.00E+00	1.08E-03	0.00E+00	5.69E-03	0.00E+00	2.49E+00	0.00E+00	1.01E-02	1.87E-01
D0945	Control	0.00E+00	4.79E-04	7.00E-03	0.00E+00	4.19E-03	3.99E+00	0.00E+00	1.11E-03	3.96E-02
D1174	Control	0.00E+00	9.33E-04	0.00E+00	3.43E-03	1.30E-02	7.21E+00	6.53E-03	0.00E+00	1.66E-02
D2054	Control	1.55E-03	0.00E+00	0.00E+00	1.22E-02	0.00E+00	2.07E+00	0.00E+00	1.28E-02	3.90E-01
D2062	Control	2.39E-05	0.00E+00	6.04E-02	2.99E-02	0.00E+00	4.94E+00	0.00E+00	9.95E-03	0.00E+00
D2079	Control	2.73E-02	0.00E+00	1.81E-03	1.17E-02	0.00E+00	3.38E+01	7.87E-02	0.00E+00	5.91E+00

Metadata

Normalized binned data



# Data Normalization, Transformation, and Scaling



## Data Normalization

- Normalization reduces the sample to sample variability due to differences in sample concentrations—particularly important when the matrix is urine
  - Normalization to total intensity is the most common method
    - For each sample, divide the individual bin integral by the total integrated intensity
  - Other Methods
    - Normalize to a peak that is always present in the same concentration, for example normalizing to creatinine
    - Probabilistic quotient normalization
    - Quantile and cubic spline normalization



## Centering, Scaling, and Transformations

**I** Centering  $\tilde{x}_{ij} = x_{ij} - \bar{x}_i$

**II** Autoscaling  $\tilde{x}_{ij} = \frac{x_{ij} - \bar{x}_i}{s_i}$

Range scaling  $\tilde{x}_{ij} = \frac{x_{ij} - \bar{x}_i}{(x_{i_{\max}} - x_{i_{\min}})}$

Pareto scaling  $\tilde{x}_{ij} = \frac{x_{ij} - \bar{x}_i}{\sqrt{s_i}}$

Vast scaling  $\tilde{x}_{ij} = \frac{(x_{ij} - \bar{x}_i)}{s_i} \cdot \frac{\bar{x}_i}{s_i}$

Level scaling  $\tilde{x}_{ij} = \frac{x_{ij} - \bar{x}_i}{\bar{x}_i}$

**III** Log transformation  $\tilde{x}_{ij} = 10 \log(x_{ij})$

$$\hat{x}_{ij} = \tilde{x}_{ij} - \bar{\tilde{x}}_i$$

Power transformation

$$\tilde{x}_{ij} = \sqrt{(x_{ij})}$$

$$\hat{x}_{ij} = \tilde{x}_{ij} - \bar{\tilde{x}}_i$$

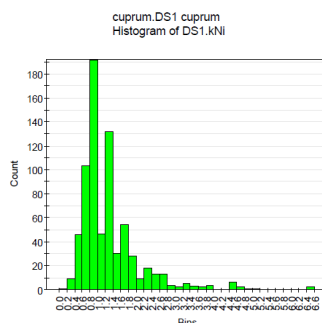
Analysis results vary depending on the scaling/ transformation methods used.

Van den Berg et al 1006, BMC Genomics, 7, 142

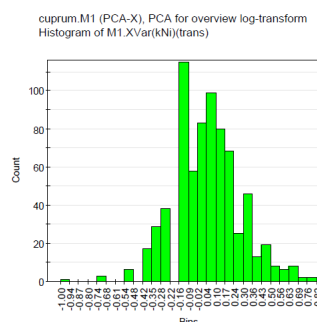


## Data Transformation

- Before transformation
  - skew distribution



- After log-transformation
  - More close to normal distribution



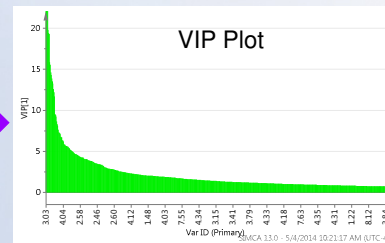
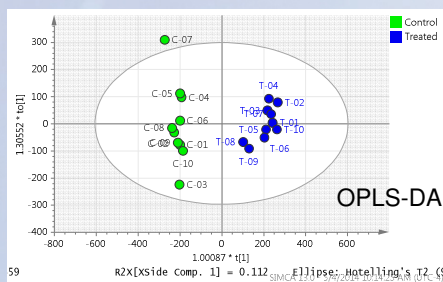
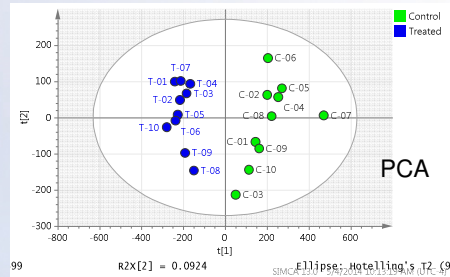
Susan Wicklund, Multivariate data analysis for omics, Sept 2-3 2008,  
Umetrics training

## Scaling

- Unit variance (autoscaling) divides the bin intensity by the standard deviation
  - May increase your baseline noise
  - Dimensionless value after scaling
- Pareto scaling divides the bin intensity by the square root of the standard deviation
  - Not dimensionless after scaling
- For NMR data, centering with pareto scaling is commonly used

# Multivariate Data Analysis and Other Statistical Analyses

- Mean centered and scaled data
- Non-supervised analysis
  - Principal component analysis (PCA)
- Supervised analysis
  - PLS-DA and OPLS-DA
- Loadings plots and VIP Plots to identify discriminatory bins
- p-Value, fold change



## STS Center

				Susan Sumner			
							
Wimal Pathmasiri NMR & GC-MS	Jim Carlson LC- and GC-MS	Jessica Gooding LC-MS	Kelly Mercier NMR	Susan McRitchie Data Analysis	Zach Acuff Biostatistics	Bob Clark Genetics	Jason Burgess Program Coordinator
							
Andrew Novokhatny NMR and QC	Aurora Cabrera LC-MS/MS	Jocelin Spruill GC-MS	Tammy Cavallo Biology and QC	Delisha Stewart Cell Biology	Ninell Mortensen Microbiology	Maria Moreno Biology	Keith Levine Metallomics
							
Yuanyuan Li LC-MS	Rod Snyder LC-MS	Sherry Black In vivo and in vitro Metabolism	Scott Watson Neurotransmitter LC/MS	Skip Gaudette Systems	Puvi Patel In vitro metabolism	Yan Lan Yueh LC-MS	Tim Fennell Metabolism
							
				Hieu Vu LC-MS			Sue Clark Administrative Support
						Courtney Whitaker LC/MS	