

Model Maintenance: When, how and how to avoid it

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Outline

- Introduction to Model Maintenance
- Identifying the need for action
- Model updating methods
- Instrument standardization methods
- Avoiding model maintenance
- Conclusions

B.M. Wise and R.T. Roginski, "Model Maintenance: the unrecognized cost in PAT and QbD," [Chemistry Today](#), Vol. 33(2), pps. 38-43, March/April, 2015.

B.M. Wise and R.T. Roginski, "A Calibration Model Maintenance Roadmap," [Proceedings of ADCHEM 2015](#), Whistler, BC, Canada, pps. 260-265, June, 2015.

Definition and Goal

- Model maintenance: The on-going servicing of multivariate models to preserve their predictive abilities.
- Goal of model maintenance: Sustain (or improve) models over time and changing conditions with the least amount of cost and effort

Why Model Maintenance?

- Numerous things can cause multivariate models to become invalid
 - samples move to a range outside original calibration
 - analyte or interferent goes beyond calibration range or occurs in unusual combination
 - new variation is introduced into the samples
 - new interferent or variation in physical parameter, *e.g.* temperature
 - a change in the sample matrix causes the relationship between analyte and measurement to change
 - change in pressure, pH, particle size, temperature
 - a change in the hardware causes the analyte-measurement relationship to change
 - instrument maintenance, fiber optic change, source replacement, etc.

Before Model Goes Online

- Develop a plan for maintenance
 - Assume that updated or new calibration models will eventually be required
 - Have a plan for how to detect the problem and what to do about it
 - Put it in the budget!
- Measure standard samples
 - Plan for registration and amplitude shifts
 - Characterize instrument in ranges important to model

Detecting Model/Data Mismatch & Performance Degradation

- Model prediction diagnostics
 - Spectral residual Q (or similar)
 - Sample distance T^2 (or similar)
- Prediction accuracy monitored via primary reference method
 - Unlikely that change not detected by diagnostics but possible
 - Risk based approach?
- Detecting that *something* has gone wrong easier than determining *what* has gone wrong.
 - unless you are monitoring via reference method!
 - ... and why aren't you?

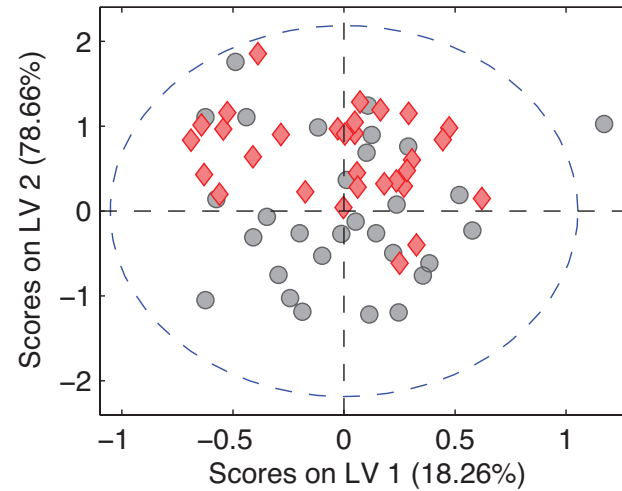
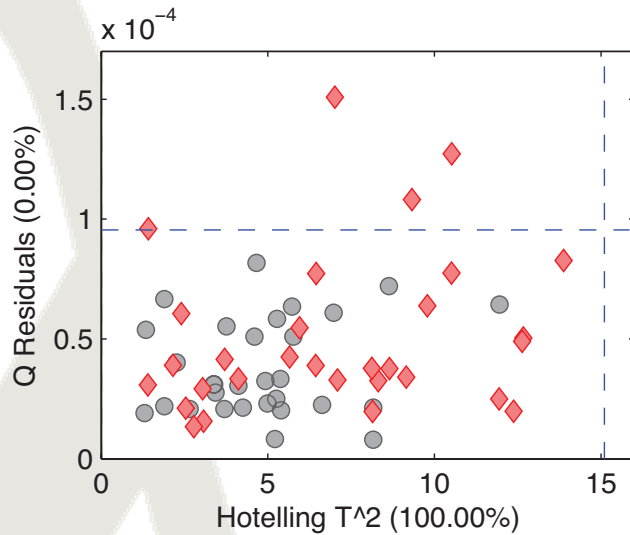
Diagnostics Limits

- By default, limits for Q and T^2 are generally provided based on some type of confidence limit
 - These limit values are statistically based
 - Meaningful limits require knowledge of the process or measurement
 - Observations are needed that represent
 - “bad” states
 - out of spec product
 - failing sensor(s) or analyzer*
- *these may be challenging to obtain

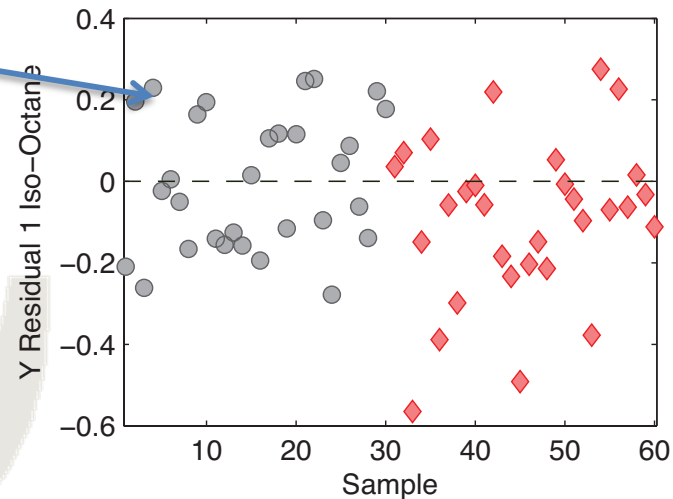
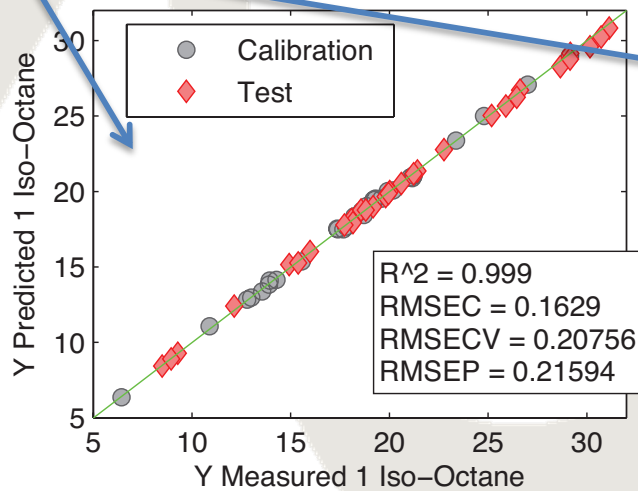
Example of detecting model/data mismatch

- Semi-synthetic example to illustrate change detection issues
- NIR measurement of iso-octane interferents (heptane, toluene, decane and eventually also xylene)
- CLS model to generate data along with a structured noise model from original data

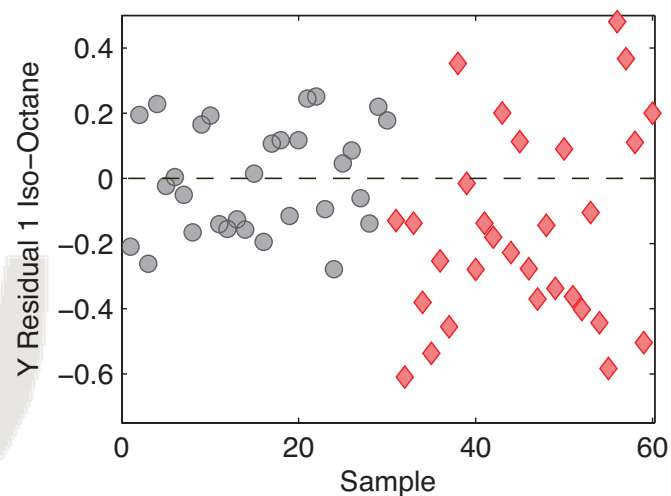
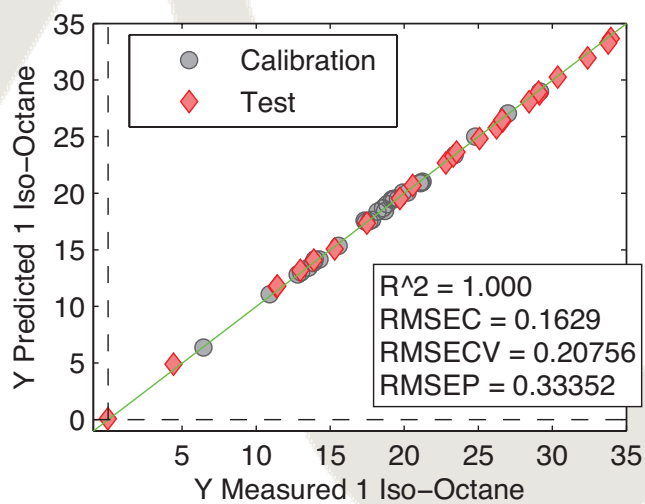
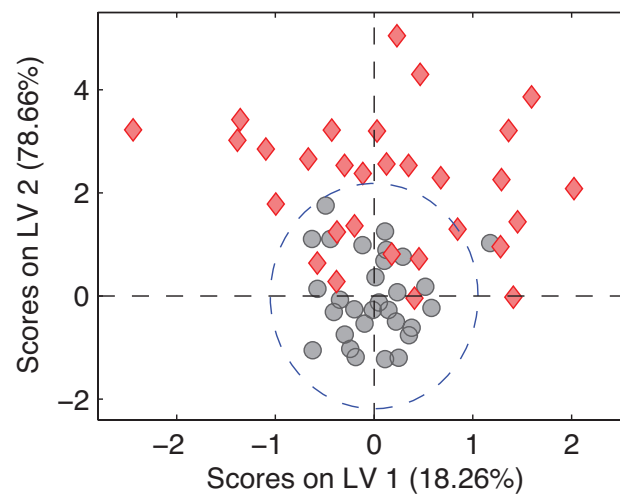
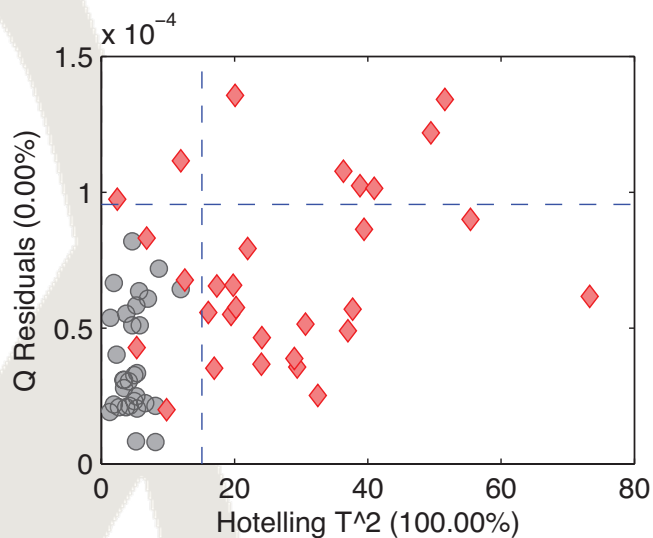
Normal Operation



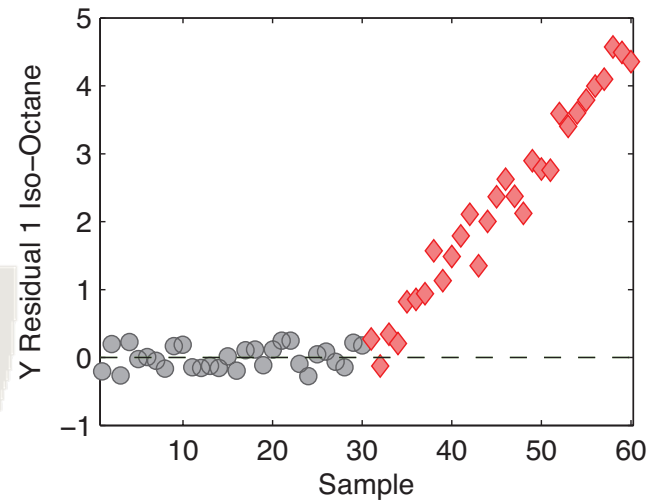
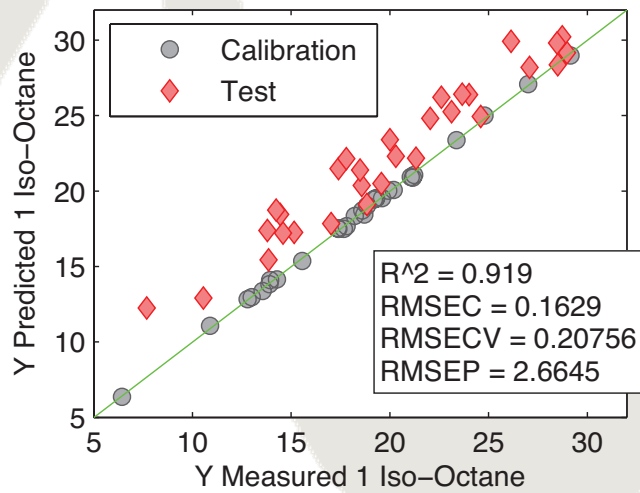
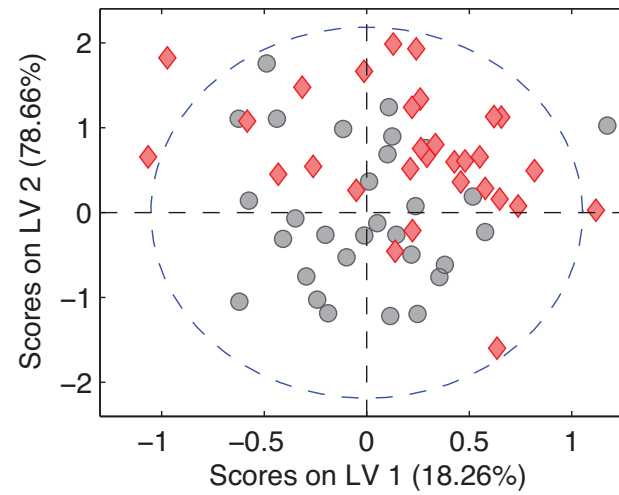
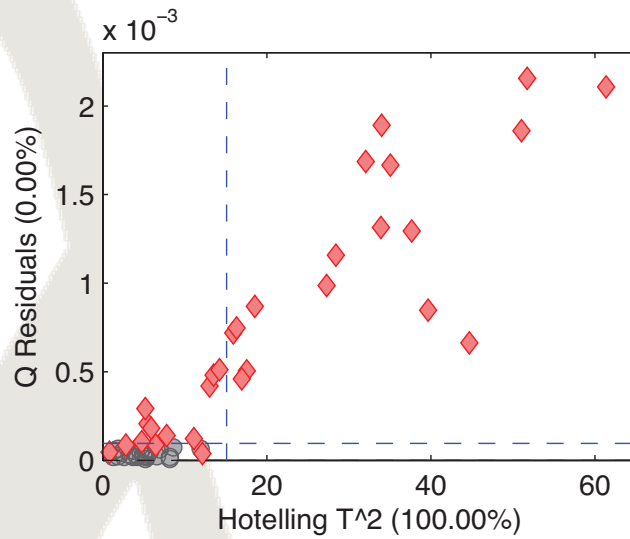
Only available from reference method



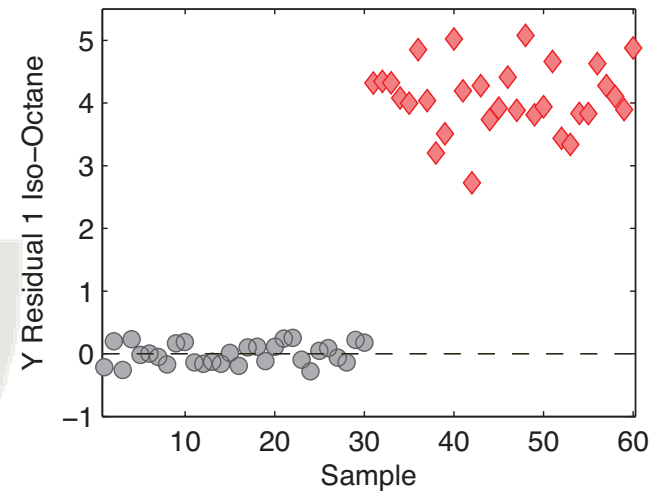
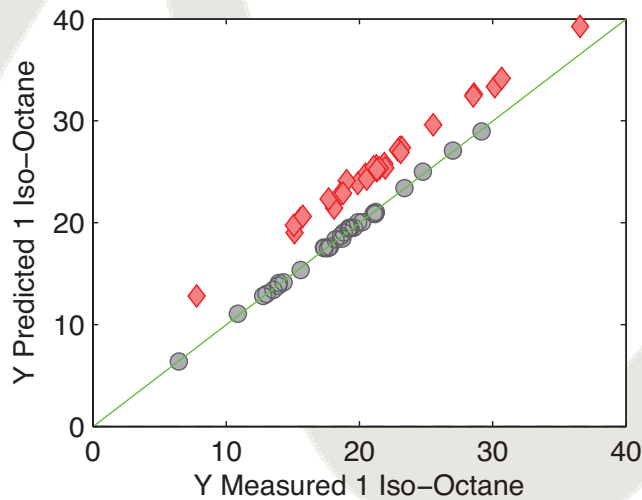
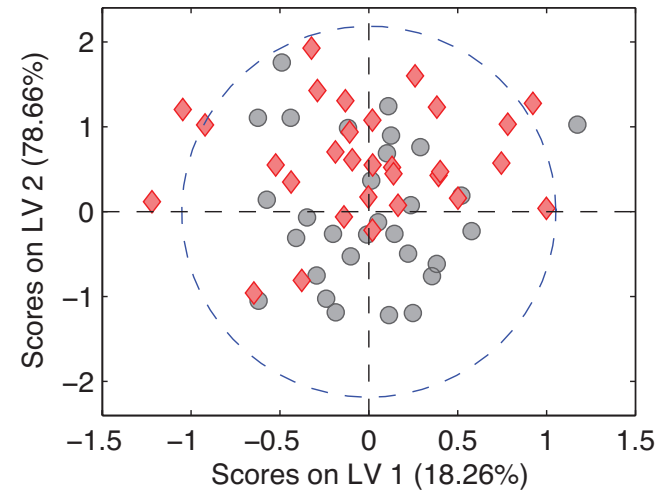
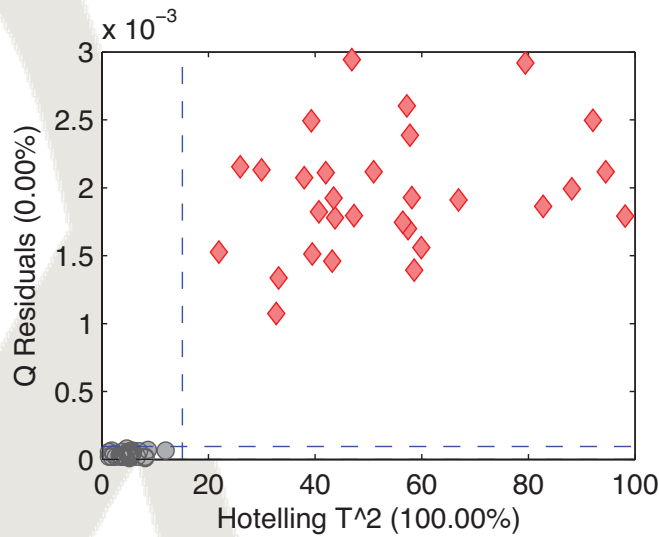
Out of Range Samples



New Interferent



Instrument Registration Shift



Determining the cause of the problem

- Might not be obvious from model diagnostics
- Measurement by reference method helpful
- Standard samples can pin down cause unambiguously
- Then what?
 - Expand calibration set
 - Slope and bias correction
 - Instrument standardization

Adding Samples to Calibration Set

- Out of range and new interferent problems can usually be solved by adding samples to existing calibration set
- Problem: might take more than a few samples to “balance” calibration
- Solution: upweighting of new samples

Slope and Bias Correction

- Simple to do
- May be appropriate for a constant shift
- Not to be used over and over!
 - Indicates problem is variable interferent or something else
- Automated model updating?

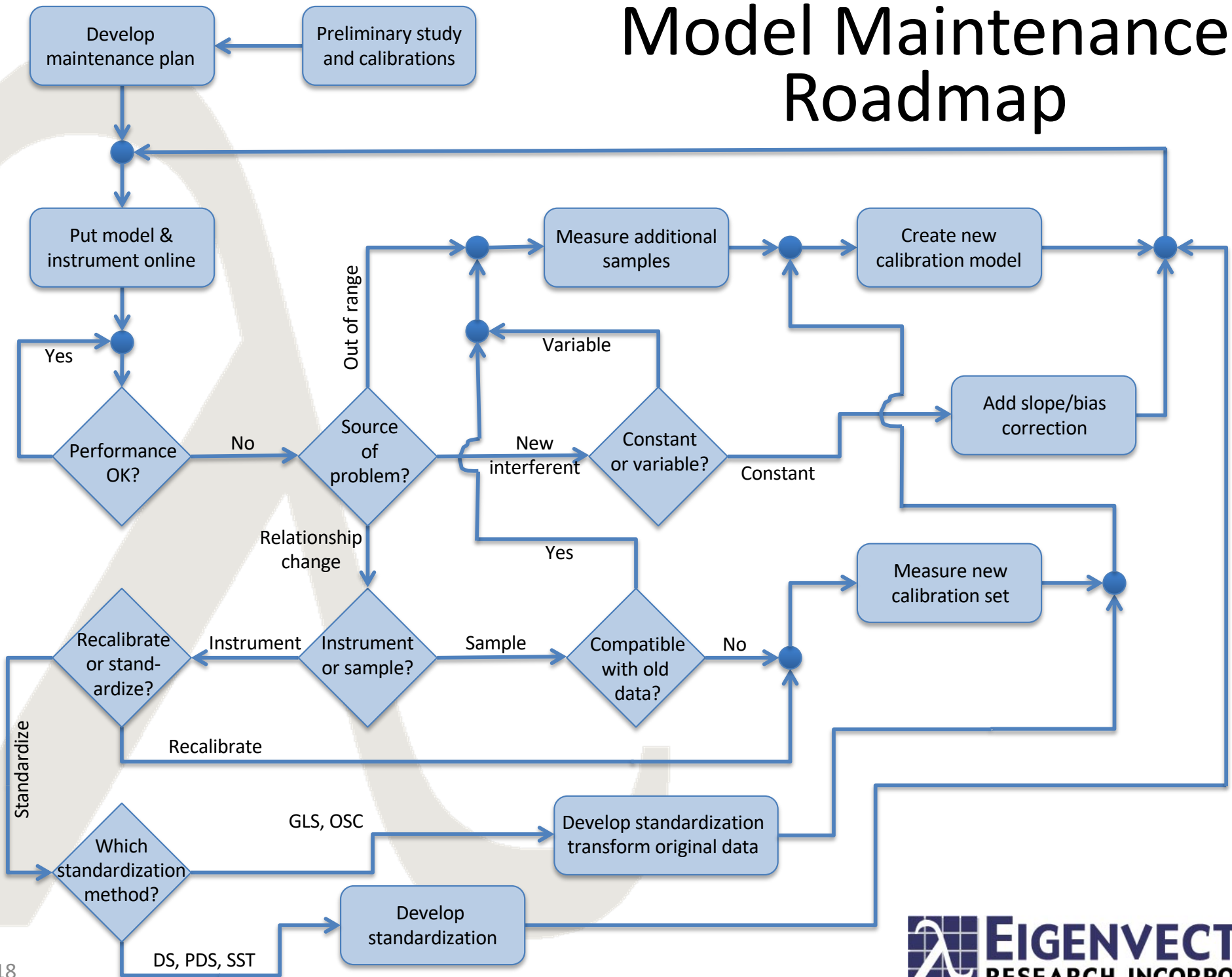
Standardization Methods

- Many methods available to estimate the response of the standard instrument from a different or changed instrument
- My favorites
 - Direct Standardization (DS)
 - Piecewise Direct Standardization (PDS)
 - Subspace Standardization Transform (SST)
 - Generalized Least Squares Preprocessing (GLS)
 - Orthogonal Signal Correction (OSC)

Putting it all together

- Have covered the pieces
 - Detecting change and performance degradation
 - Identifying the problem
 - Methods for updating/correcting models
 - Standardization methods
- How does this all fit together?

Model Maintenance Roadmap



Avoiding Model Maintenance

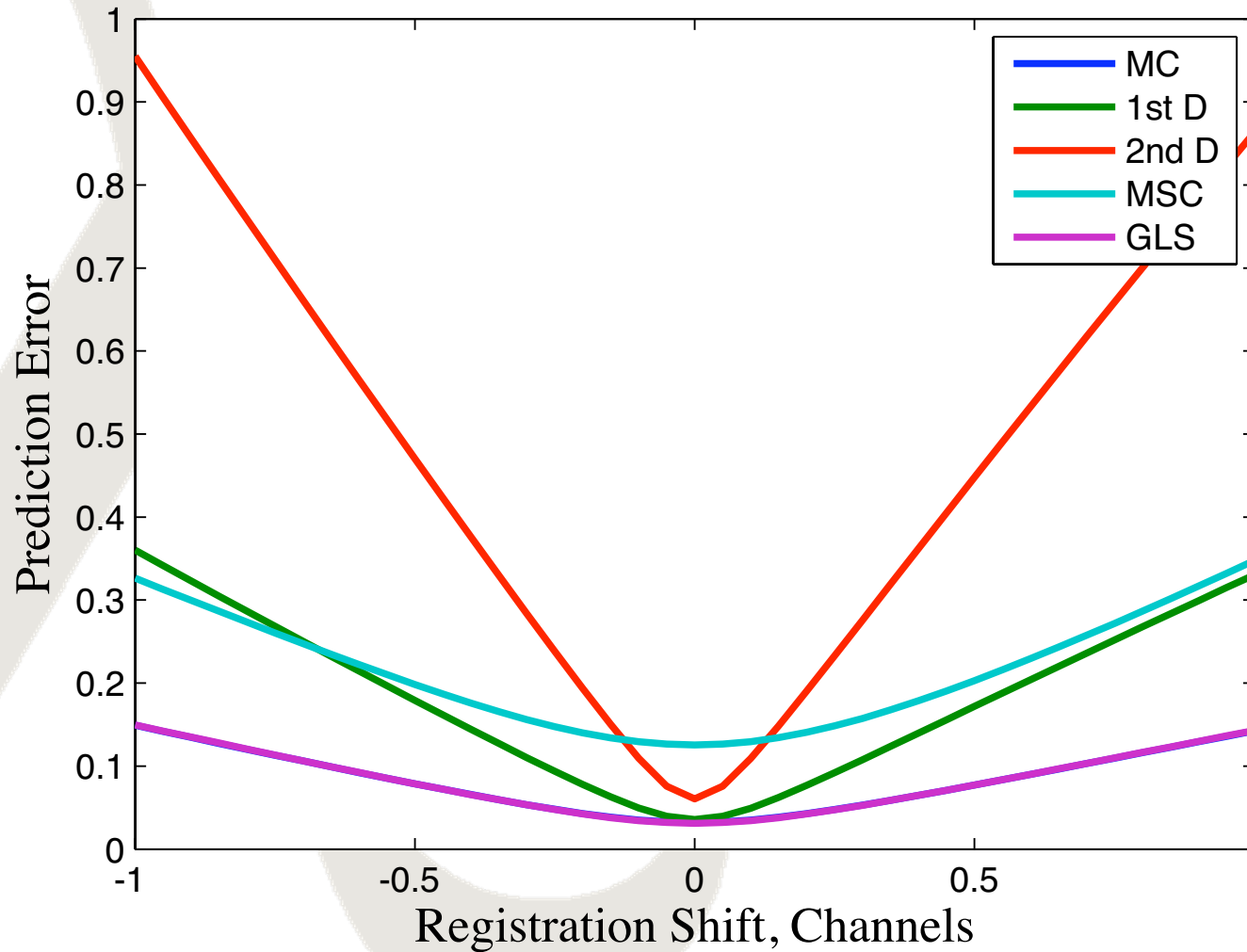
- Some models more robust to new analytes and changes in data than others
- Highly dependent on preprocessing options and number of factors in models

Robustness Tests

- Series of functions developed to test model against system changes
 - Develop model with desired preprocessing, #LVs, etc.
 - “Perturb” test data set
 - Apply calibration model to “perturbed” data
 - Look at prediction error as function of perturbations
 - Test and compare multiple models

Shift with Preprocessing

Registration Shift Errors for Corn Models



Conclusions

- Plan and budget (!) for model maintenance
- Many elements
 - Performance monitoring
 - Problem detection and identification
 - Standardization protocols
 - Remodeling and revalidating
- Model robustness testing can help minimize the need for model updating