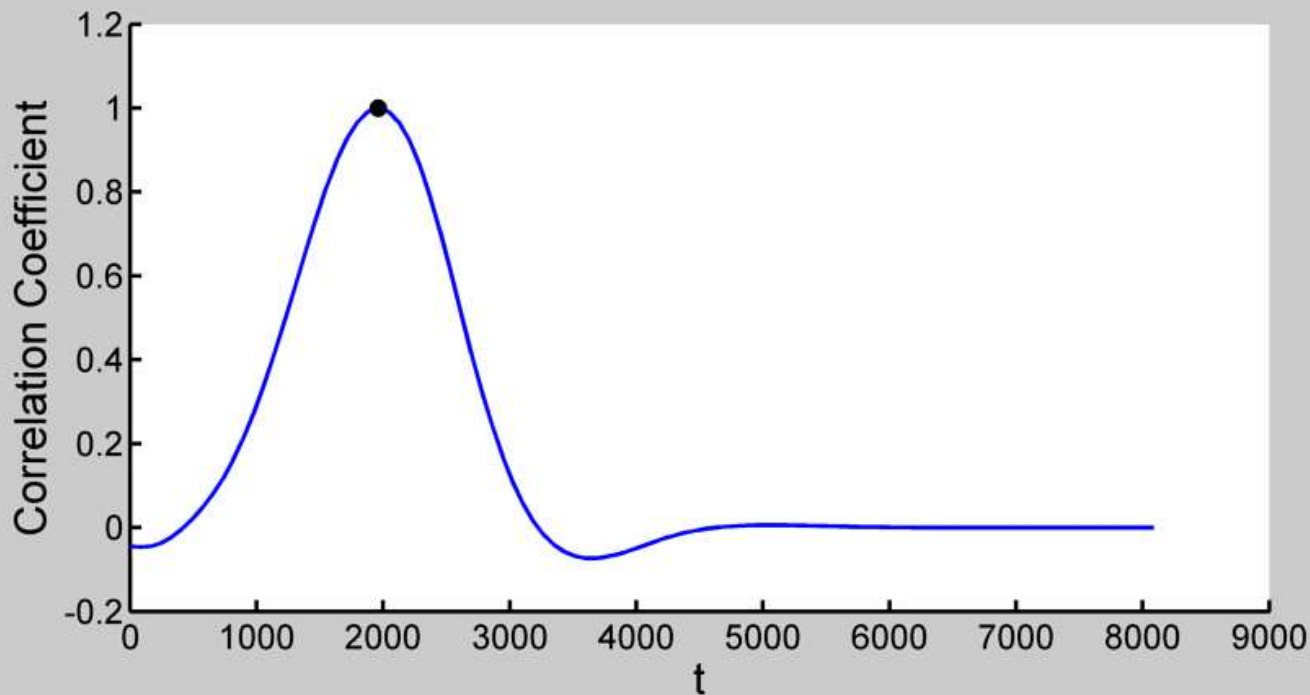


Use and abuse of weighted means/linear regression

- propagating random and systematic uncertainties

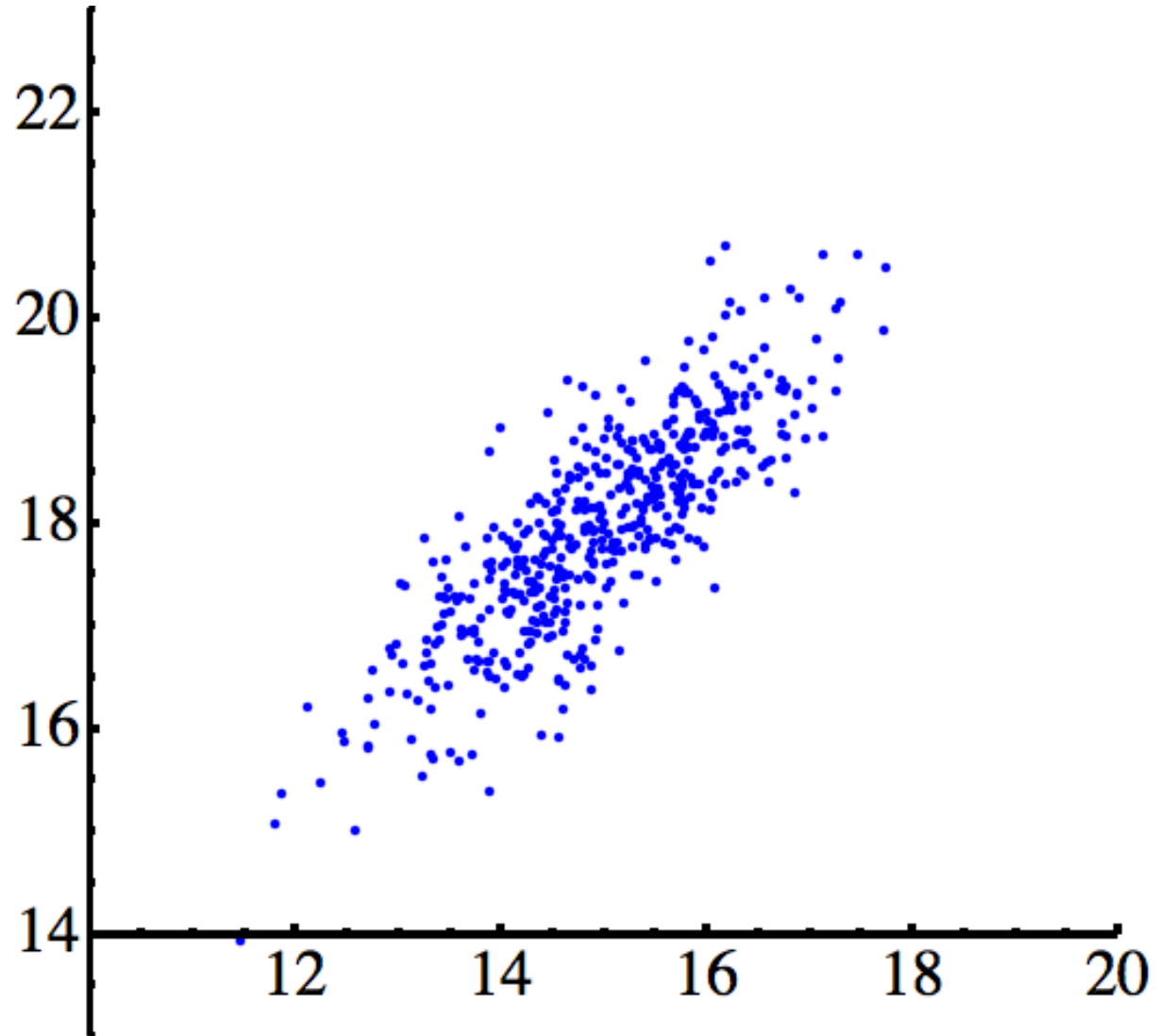


Propagating systematic uncertainties

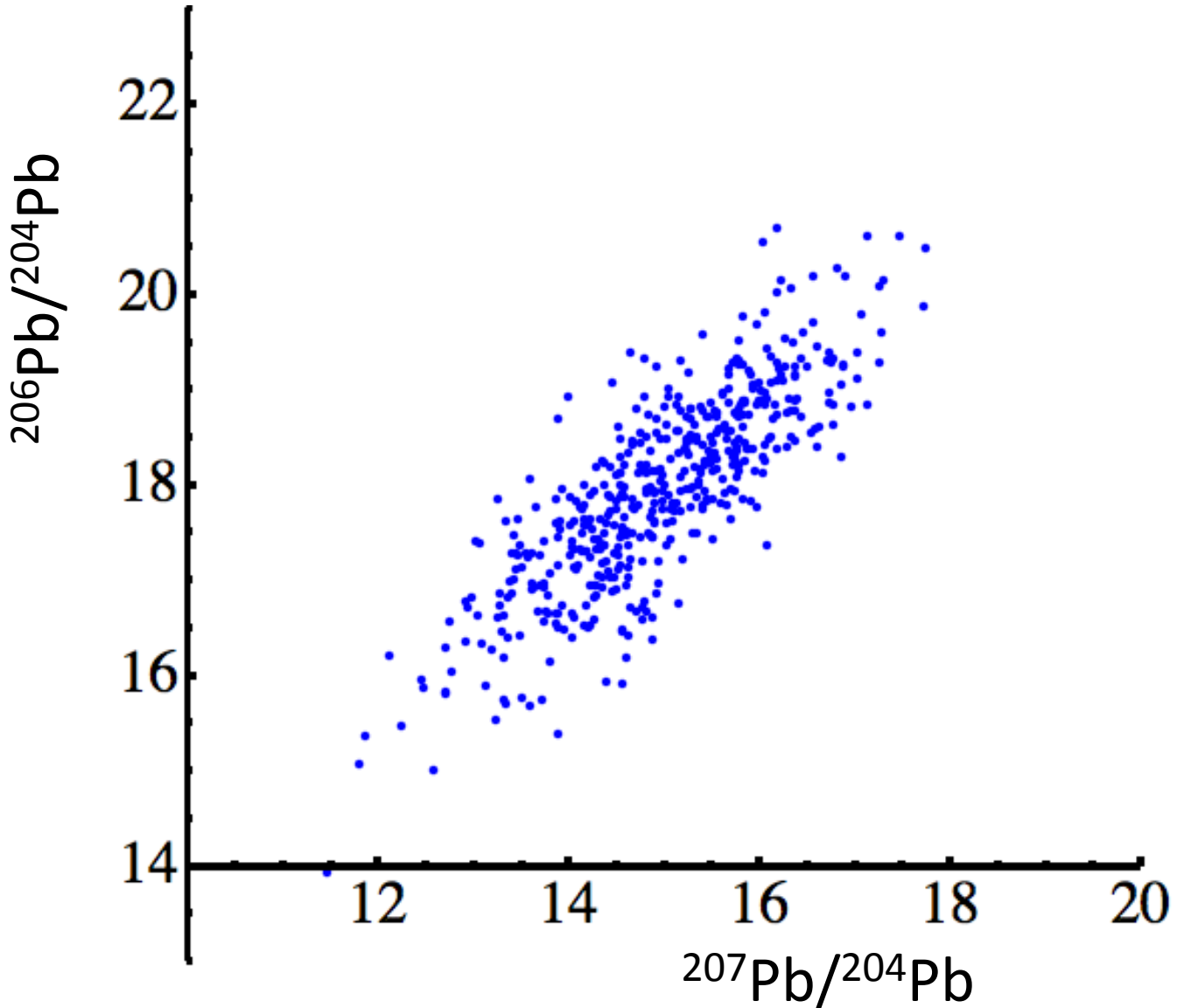
- if z is a function of x and y

$$\sigma_z^2 = \sigma_x^2 \left(\frac{dz}{dx} \right)^2 + 2\sigma_{xy}^2 \left(\frac{dz}{dx} \right) \left(\frac{dz}{dy} \right) + \sigma_y^2 \left(\frac{dz}{dy} \right)^2$$

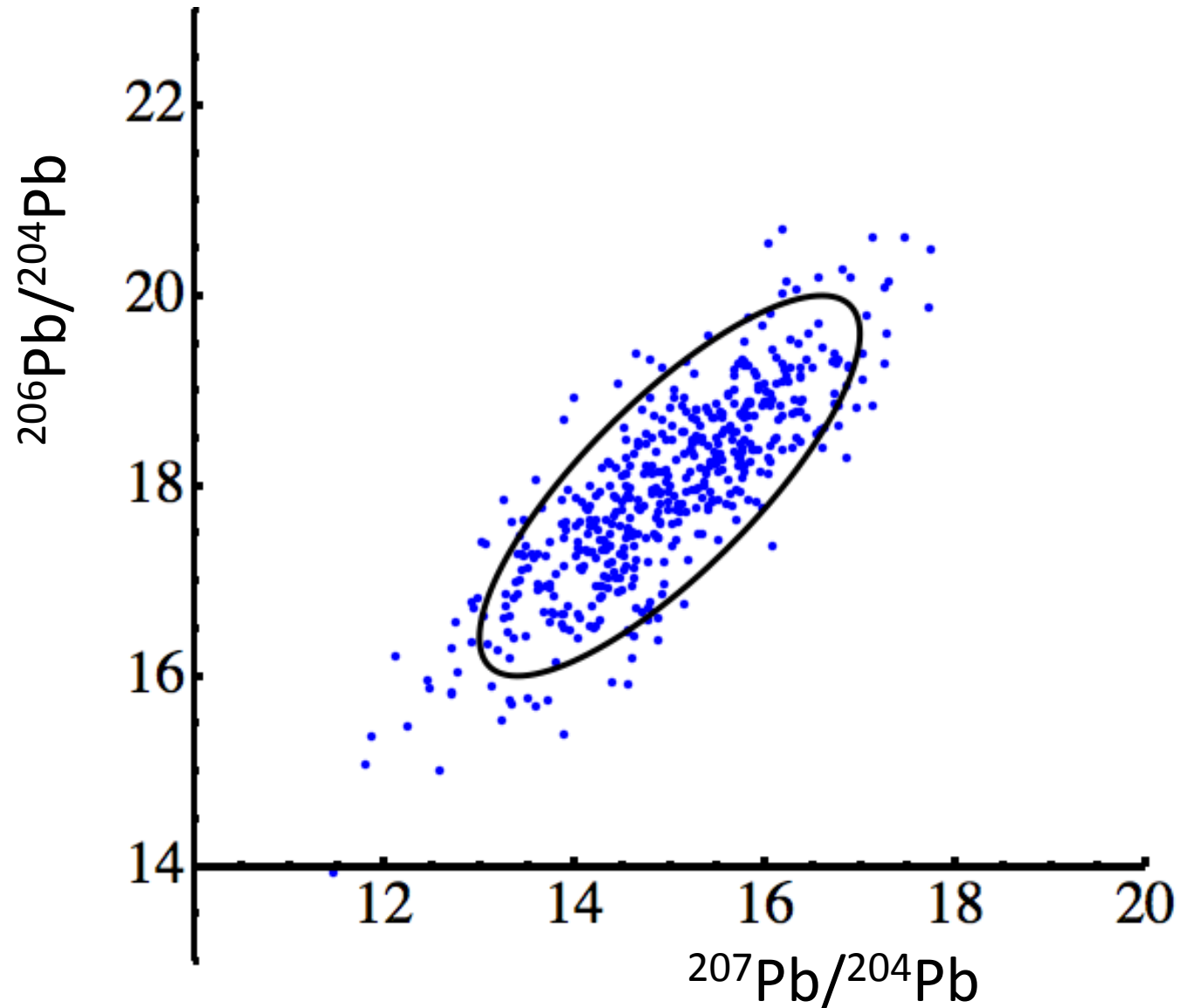
Covariance/Correlation

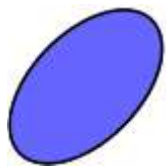


Covariance/Correlation

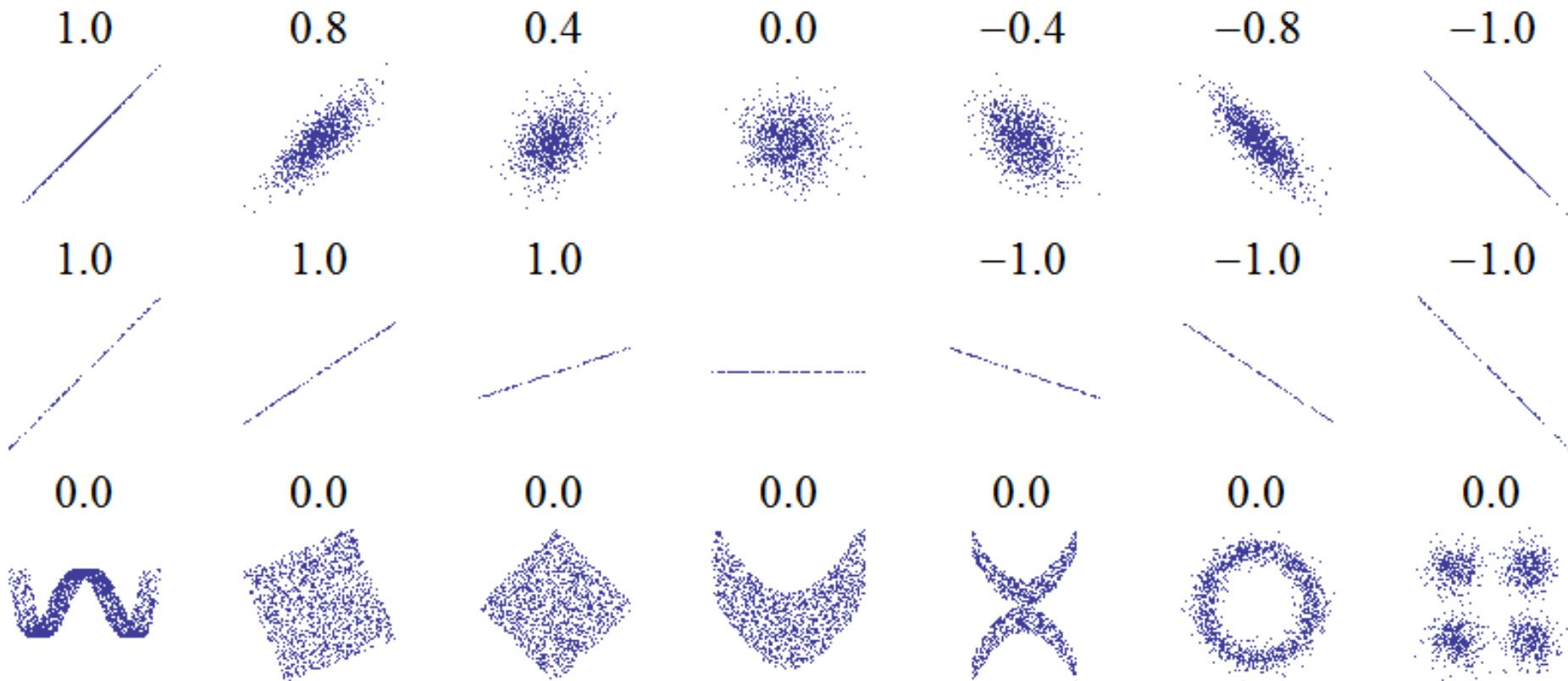


Covariance/Correlation





Other Examples



Propagating systematic uncertainties

- if z is a function of x and y

$$\sigma_z^2 = \sigma_x^2 \left(\frac{dz}{dx} \right)^2 + 2\sigma_{xy}^2 \left(\frac{dz}{dx} \right) \left(\frac{dz}{dy} \right) + \sigma_y^2 \left(\frac{dz}{dy} \right)^2$$

Propagating systematic uncertainties

- if z is a function of x and y

$$\sigma_z^2 = \sigma_x^2 \left(\frac{dz}{dx} \right)^2 + 2\sigma_{xy}^2 \left(\frac{dz}{dx} \right) \left(\frac{dz}{dy} \right) + \sigma_y^2 \left(\frac{dz}{dy} \right)^2$$

- *or*

$$\sigma_z^2 = \begin{bmatrix} \frac{dz}{dx} & \frac{dz}{dy} \end{bmatrix} \begin{bmatrix} \sigma_x^2 & \sigma_{xy}^2 \\ \sigma_{xy}^2 & \sigma_y^2 \end{bmatrix} \begin{bmatrix} \frac{dz}{dx} \\ \frac{dz}{dy} \end{bmatrix}$$



Covariance vs. Correlation

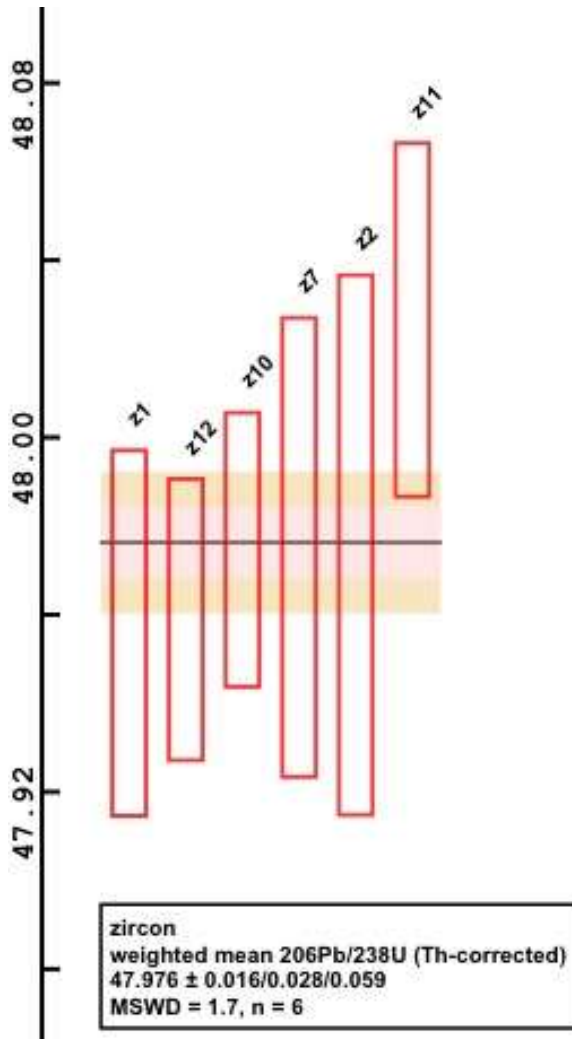
- Covariance, like variance, expresses an average sum of multiplied distances

$$\sigma_{xy}^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})$$

$$\sigma_{xy}^2 = \sigma_c^2 \left(\frac{\partial x}{\partial c} \right) \left(\frac{\partial y}{\partial c} \right)$$

$$\rho_{xy} = \frac{\sigma_{xy}^2}{\sigma_x \sigma_y} \quad [-1, 1]$$

Calculating a weighted mean



- Gives more weight to more precise analyses, un-weights less precise analyses

$$\bar{t} = \sum_{i=1}^n \alpha_i t_i = \sum_{i=1}^n \left(\frac{t_i}{\sigma_i^2} \right) / \sum_{i=1}^n \left(\frac{1}{\sigma_i^2} \right)$$

- But there is no room here for systematic uncertainties, from sample-standard bracketing, standard ICs, decay constants...

Systematic uncertainties are covariance

$$\bar{t} = \alpha_1 t_1 + \alpha_2 t_2 \quad \alpha_1 + \alpha_2 = 1$$

$$\sigma_{\bar{t}}^2 = \begin{bmatrix} \frac{d\bar{t}}{dt_1} & \frac{d\bar{t}}{dt_2} \end{bmatrix} \begin{bmatrix} \sigma_{t_1}^2 & \sigma_{t_1 t_2}^2 \\ \sigma_{t_1 t_2}^2 & \sigma_{t_2}^2 \end{bmatrix} \begin{bmatrix} \frac{d\bar{t}}{dt_1} \\ \frac{d\bar{t}}{dt_2} \end{bmatrix}$$

- What are the weights that minimize the uncertainty in the weighted mean?

Systematic uncertainties are covariance

$$\bar{t} = \alpha_1 t_1 + \alpha_2 t_2 \quad \alpha_1 + \alpha_2 = 1$$

$$\sigma_{\bar{t}}^2 = \begin{bmatrix} \frac{d\bar{t}}{dt_1} & \frac{d\bar{t}}{dt_2} \end{bmatrix} \begin{bmatrix} \sigma_{t_1}^2 & \sigma_{t_1 t_2}^2 \\ \sigma_{t_1 t_2}^2 & \sigma_{t_2}^2 \end{bmatrix} \begin{bmatrix} \frac{d\bar{t}}{dt_1} \\ \frac{d\bar{t}}{dt_2} \end{bmatrix}$$

$$\alpha = \Sigma^{-1} \mathbf{1} / (\mathbf{1}^T \Sigma^{-1} \mathbf{1})$$

Systematic uncertainties are covariance

$$\bar{t} = \alpha_1 t_1 + \alpha_2 t_2 \quad \alpha_1 + \alpha_2 = 1$$

$$\sigma_{\bar{t}}^2 = \begin{bmatrix} \frac{d\bar{t}}{dt_1} & \frac{d\bar{t}}{dt_2} \end{bmatrix} \begin{bmatrix} \sigma_{t_1}^2 & \sigma_{t_1 t_2}^2 \\ \sigma_{t_1 t_2}^2 & \sigma_{t_2}^2 \end{bmatrix} \begin{bmatrix} \frac{d\bar{t}}{dt_1} \\ \frac{d\bar{t}}{dt_2} \end{bmatrix}$$

$$\bar{t} = \mathbf{1}^T \boldsymbol{\Sigma}^{-1} \mathbf{t} / (\mathbf{1}^T \boldsymbol{\Sigma}^{-1} \mathbf{1})$$

In general:

- Random uncertainties (analytical; random variability from analysis to analysis) can be decreased with more analyses. These appear only on the diagonal of the covariance matrix
- Systematic uncertainties are usually constant biases and/or biases that vary predictably. These contribute to both the on- and off-diagonal terms of the covariance matrix.

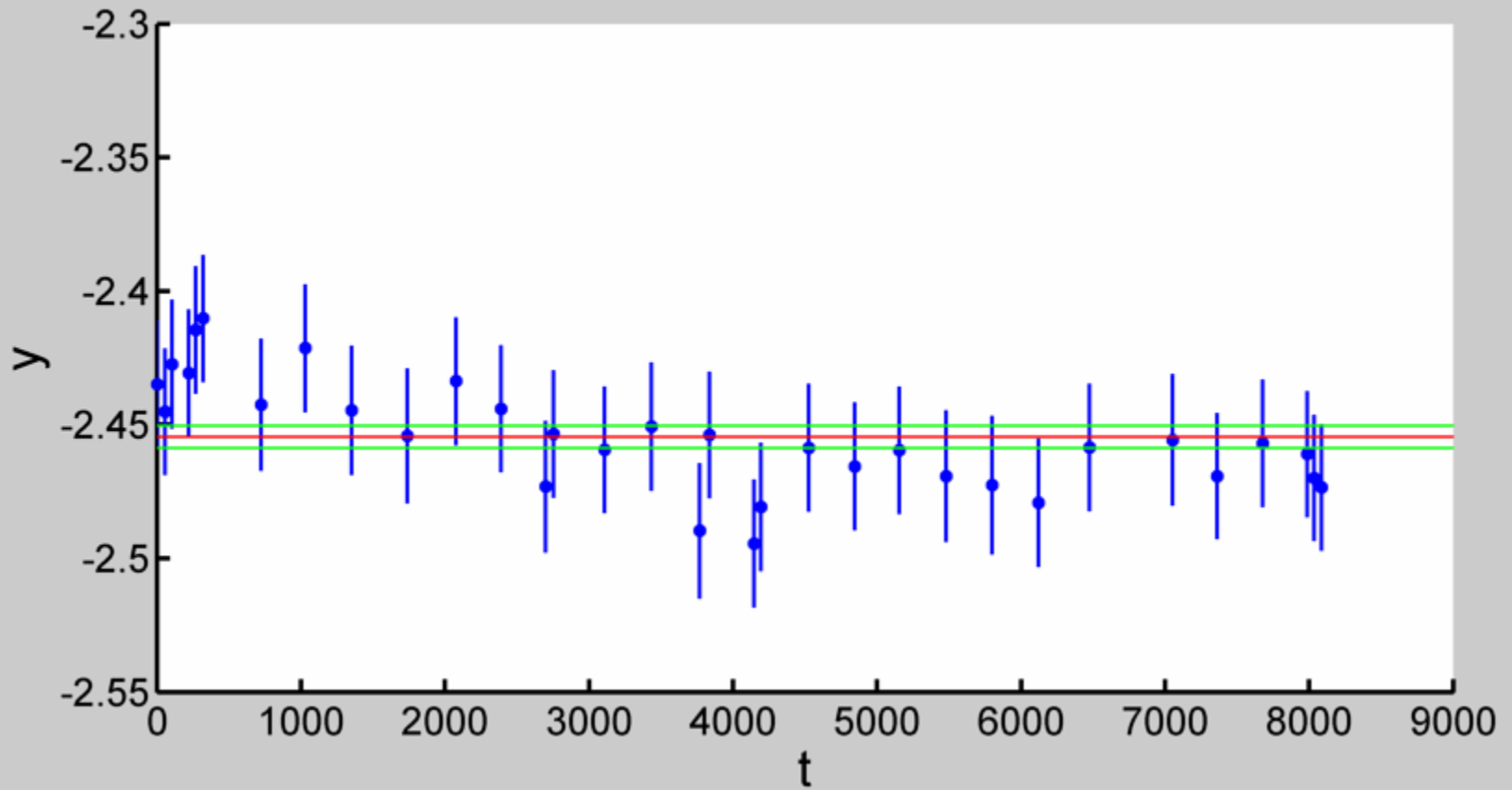
Sample-Standard Bracketing

- Central problem:
 - When we measure the standard, we don't get the true value.
 - Some normalization factor is needed to make up the difference.
 - Often times in IRMS, this takes the form an equation like $X_{\text{true}} = A X_{\text{meas}}$, where A is the fudge factor that will transform a measured value to a true value, and X is an isotope ratio.
 - A might or might not be time-dependent.

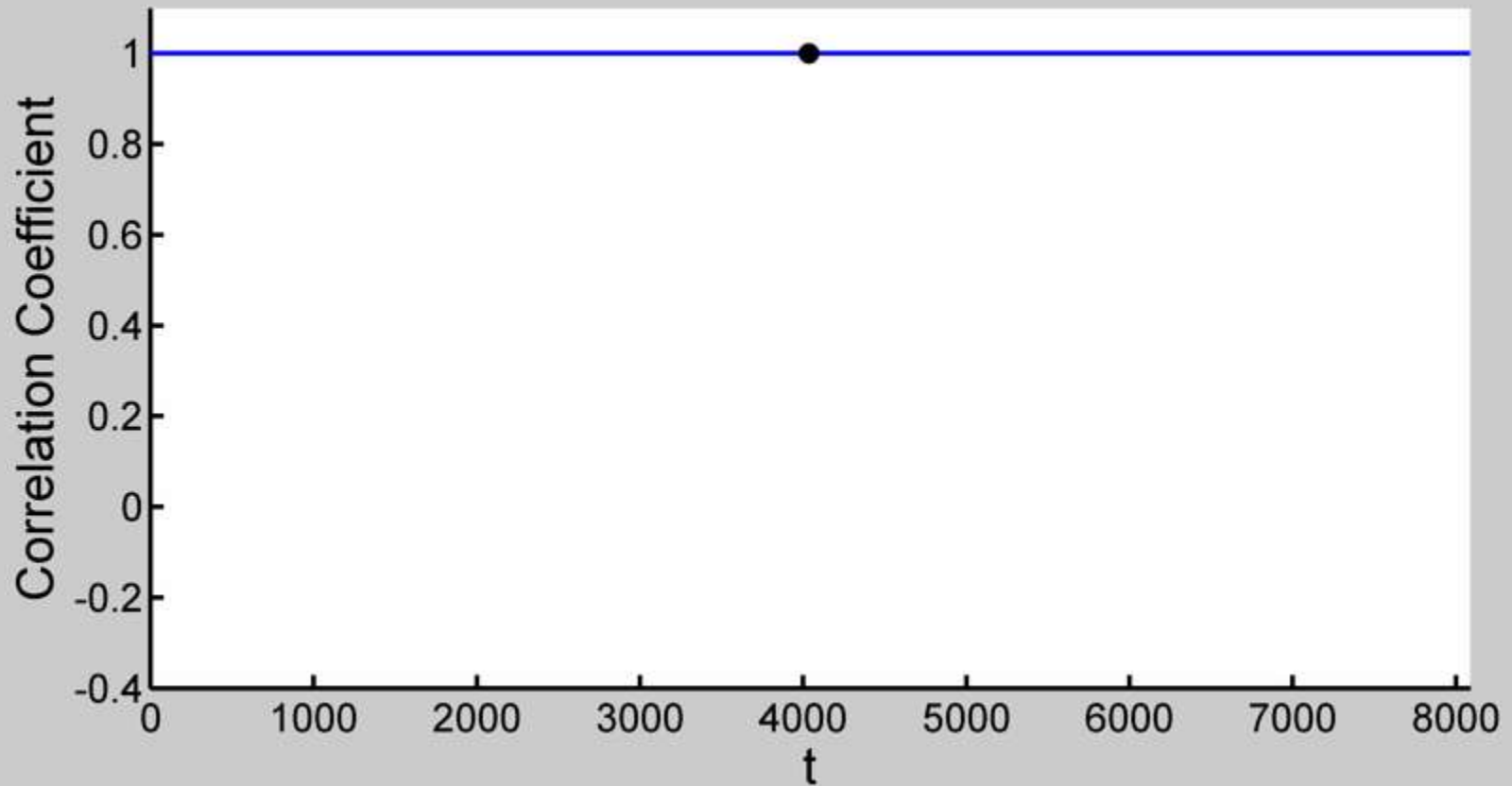
Sample-Standard Bracketing

- **Solution:** periodically measure standards, calculate A , test for time-dependence. Fit some kind of function to the values of A through time, interpolate to find the predicted values for the unknowns, propagate uncertainties
- This is all straightforward... what's the catch?
 - The uncertainties in the interpolated unknowns are not independent—they are affected by the standards. The structure of this dependency is important, and can be determined analytically.

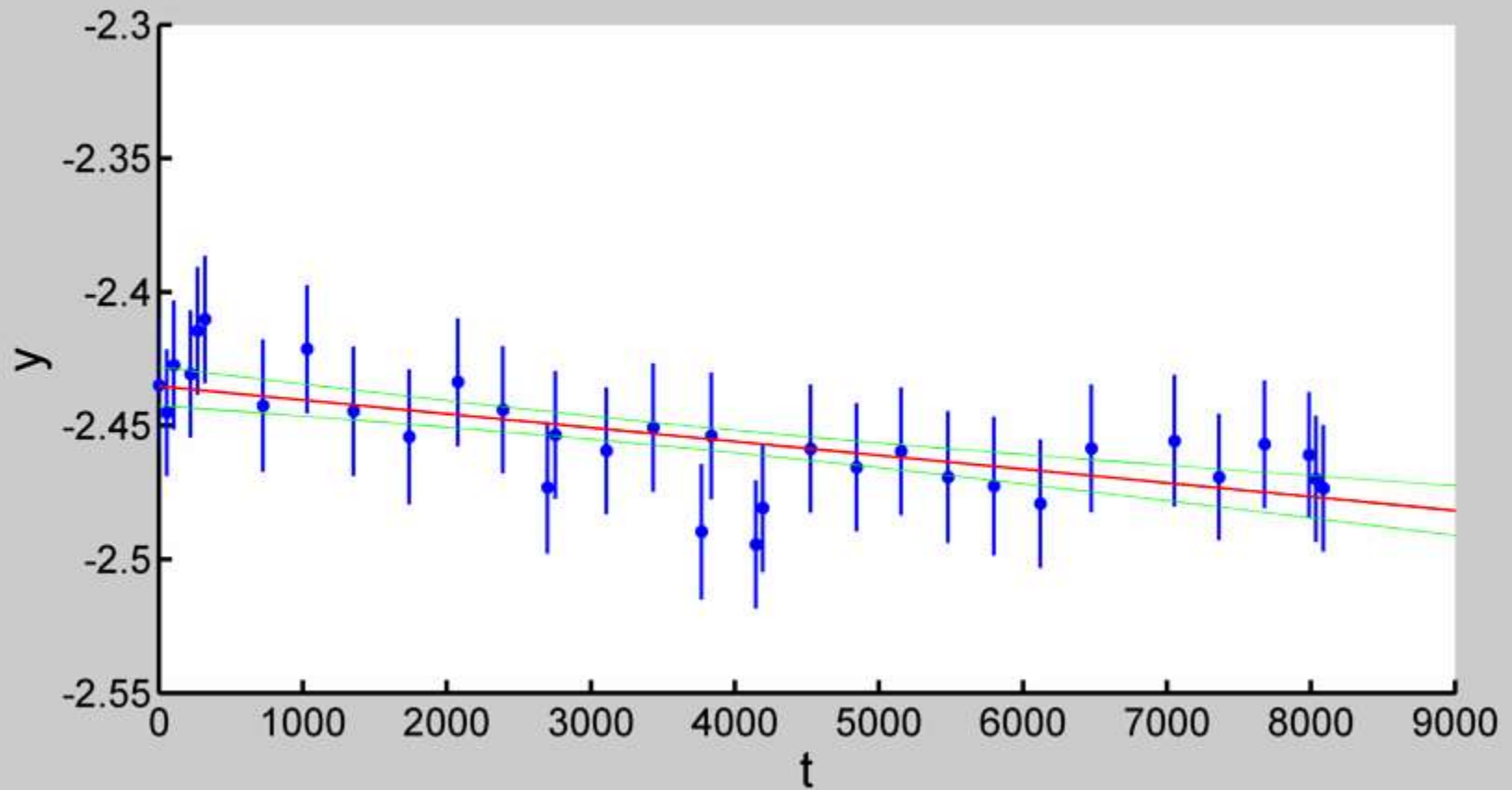
For a mean:



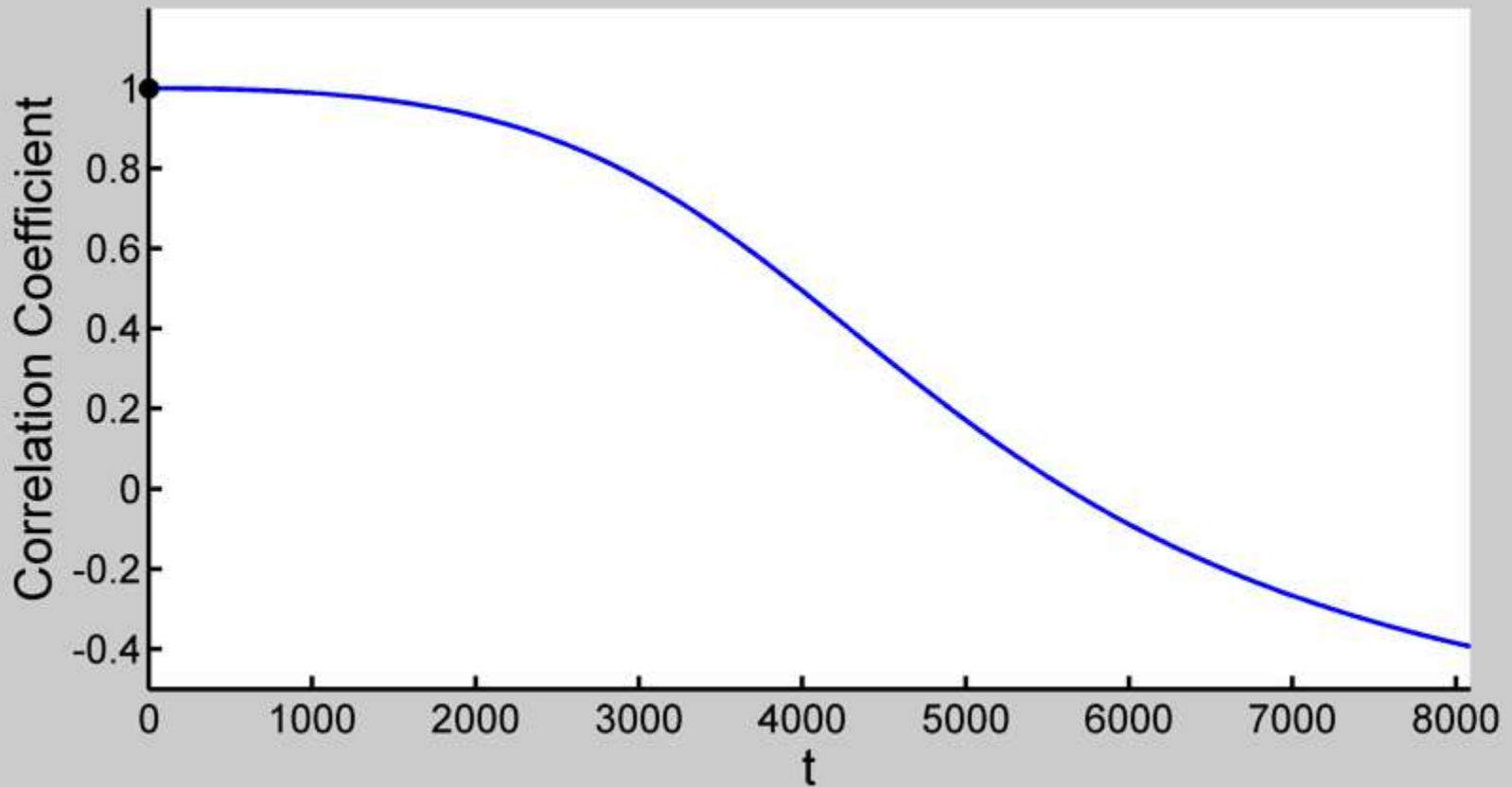
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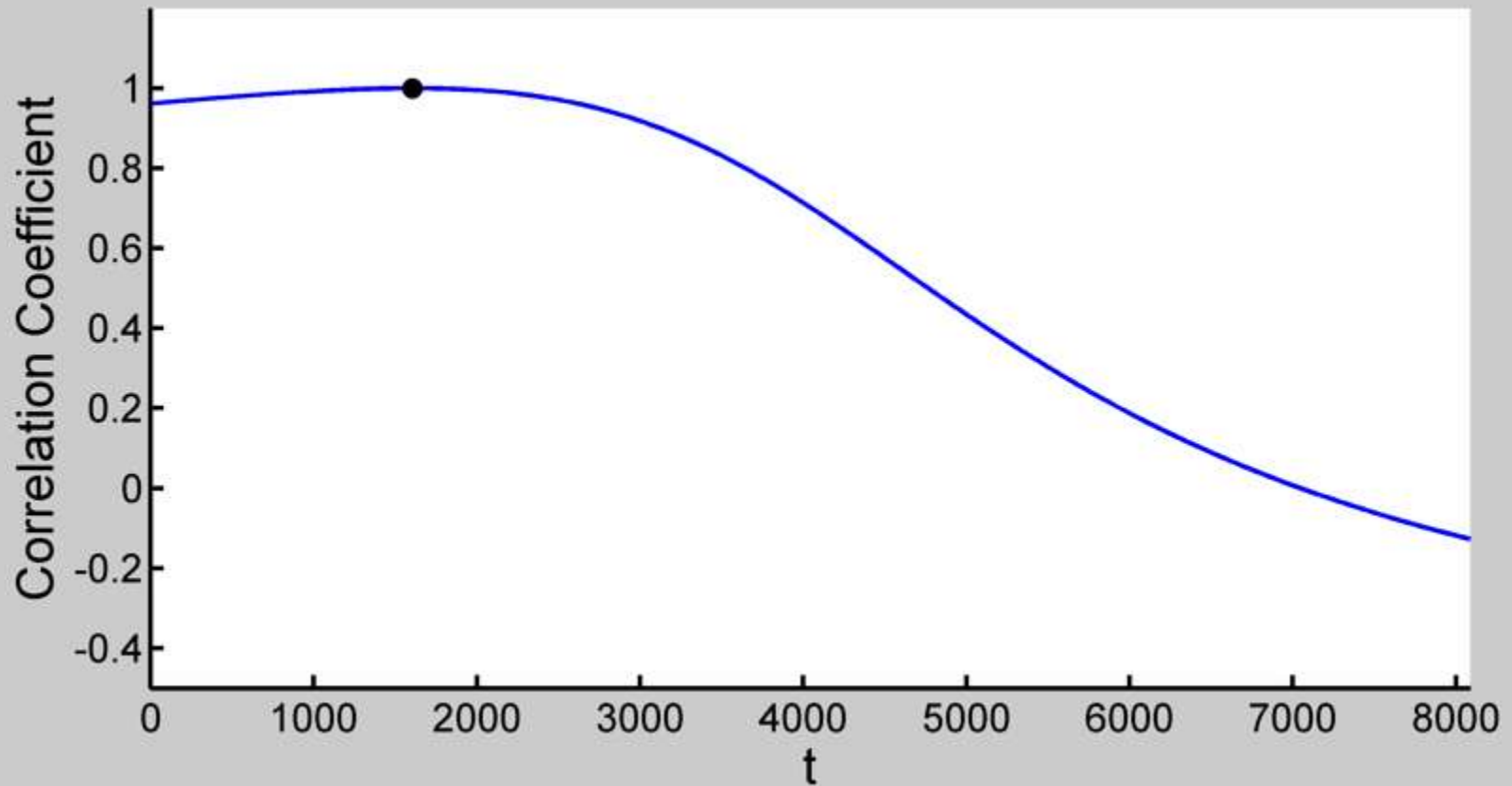
For a line:



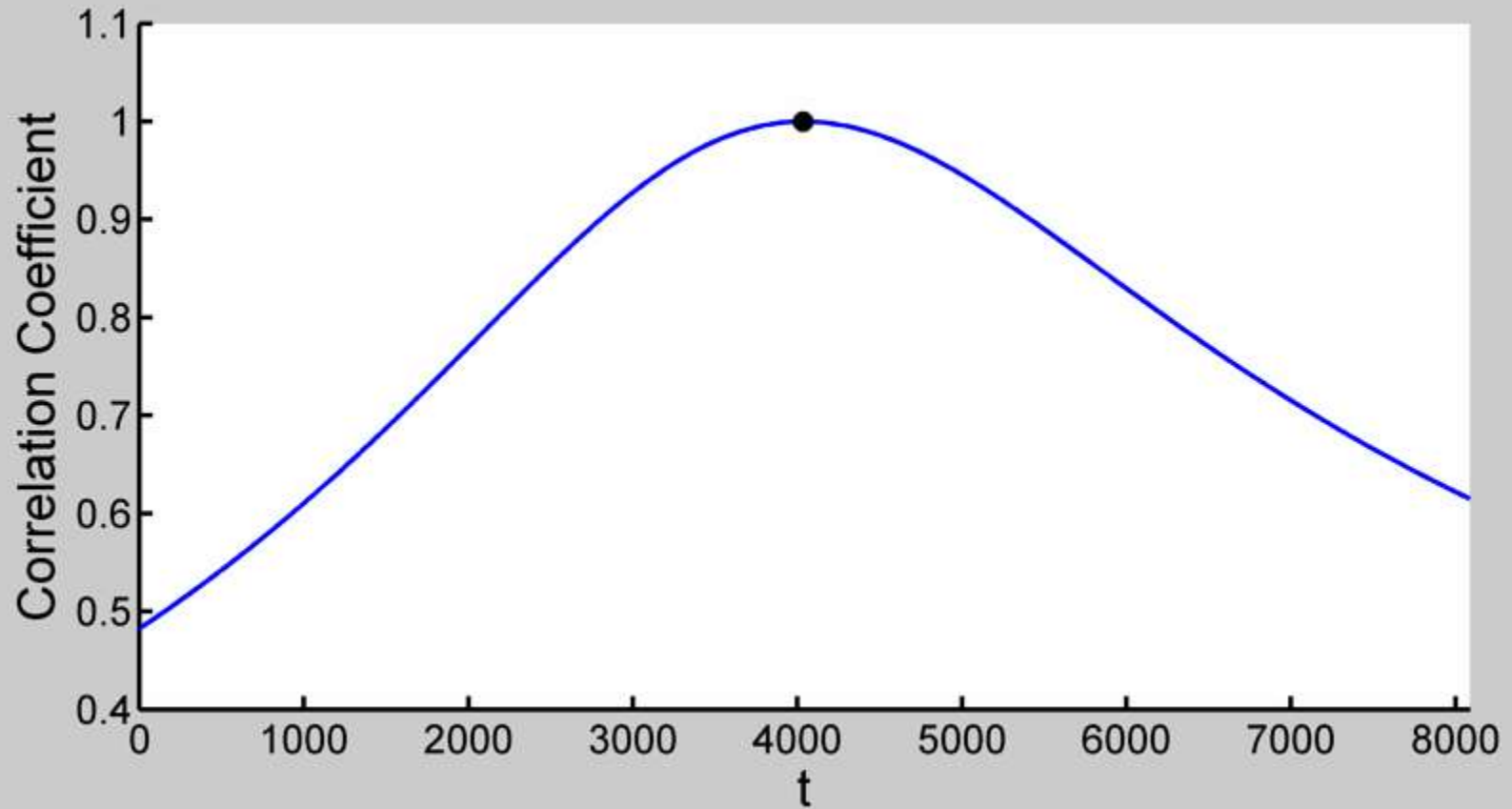
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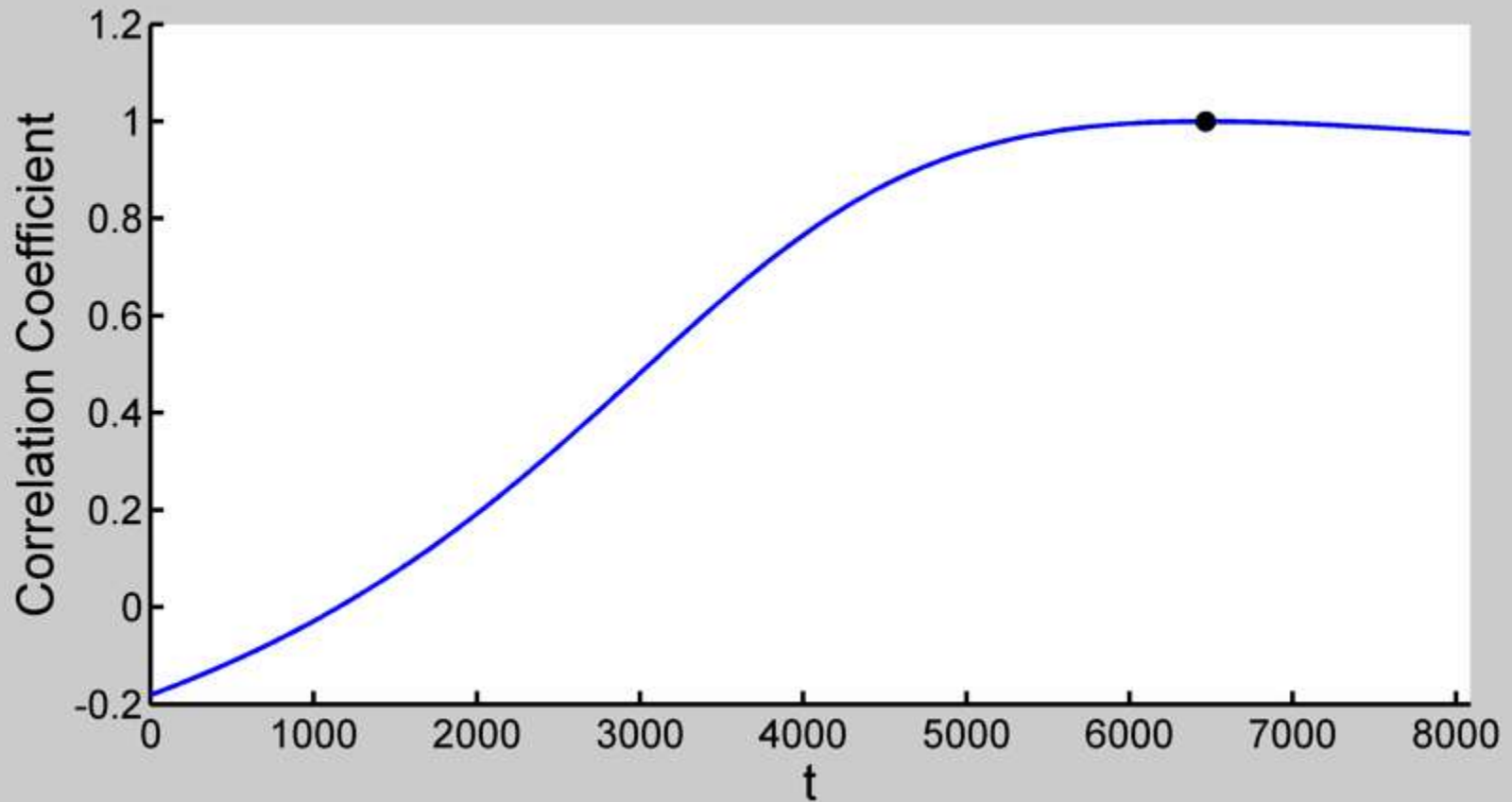
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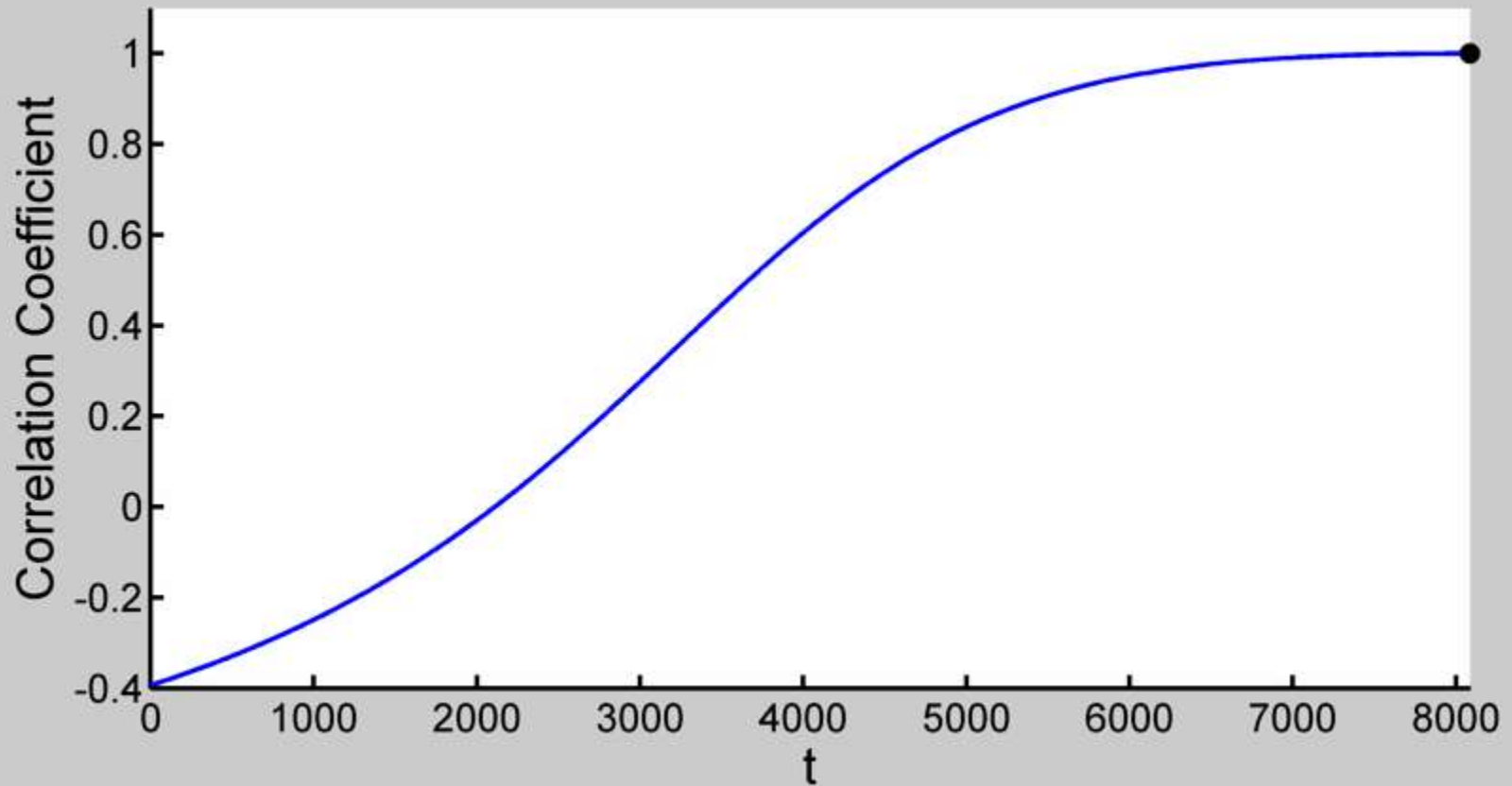
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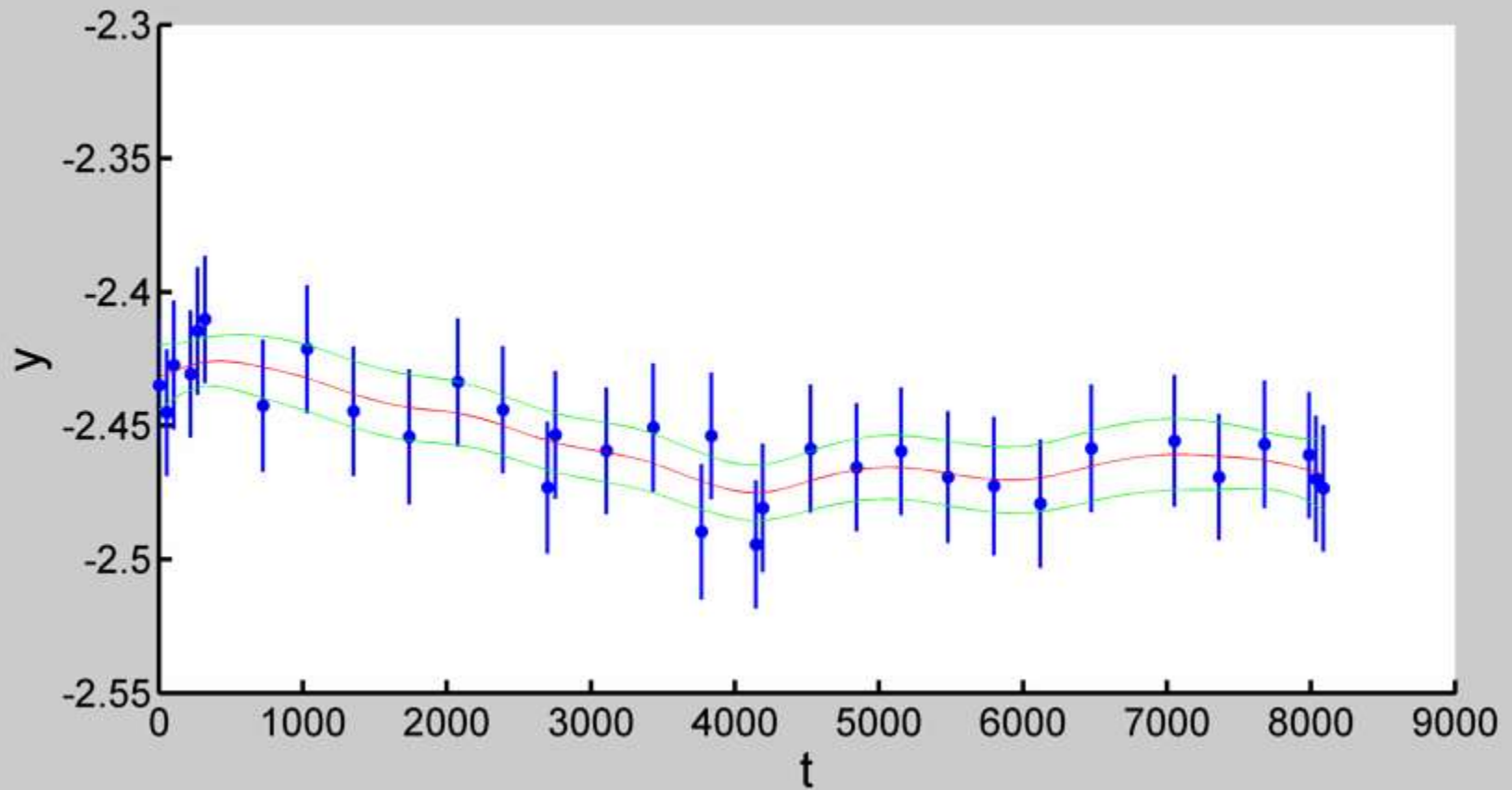
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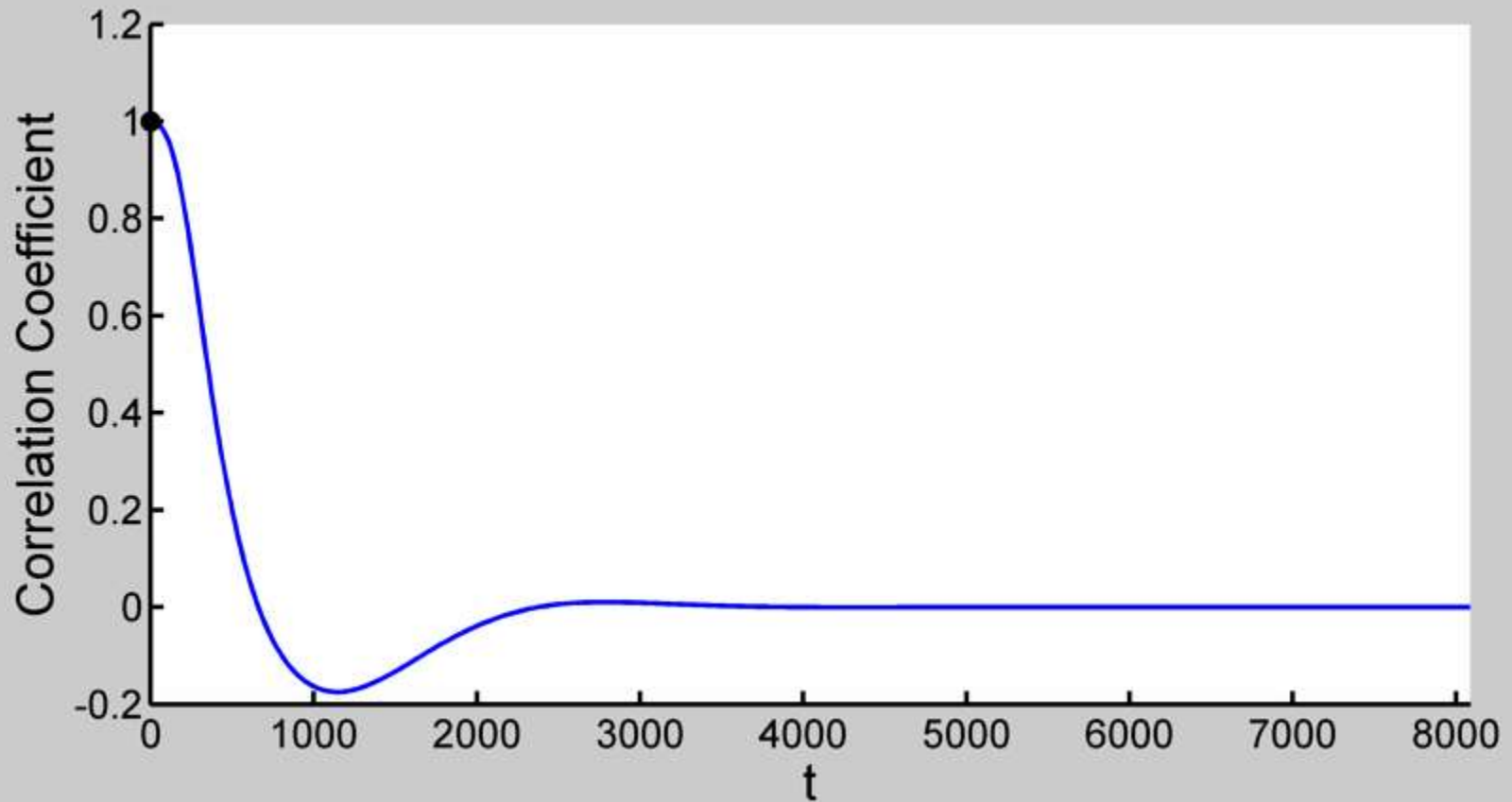
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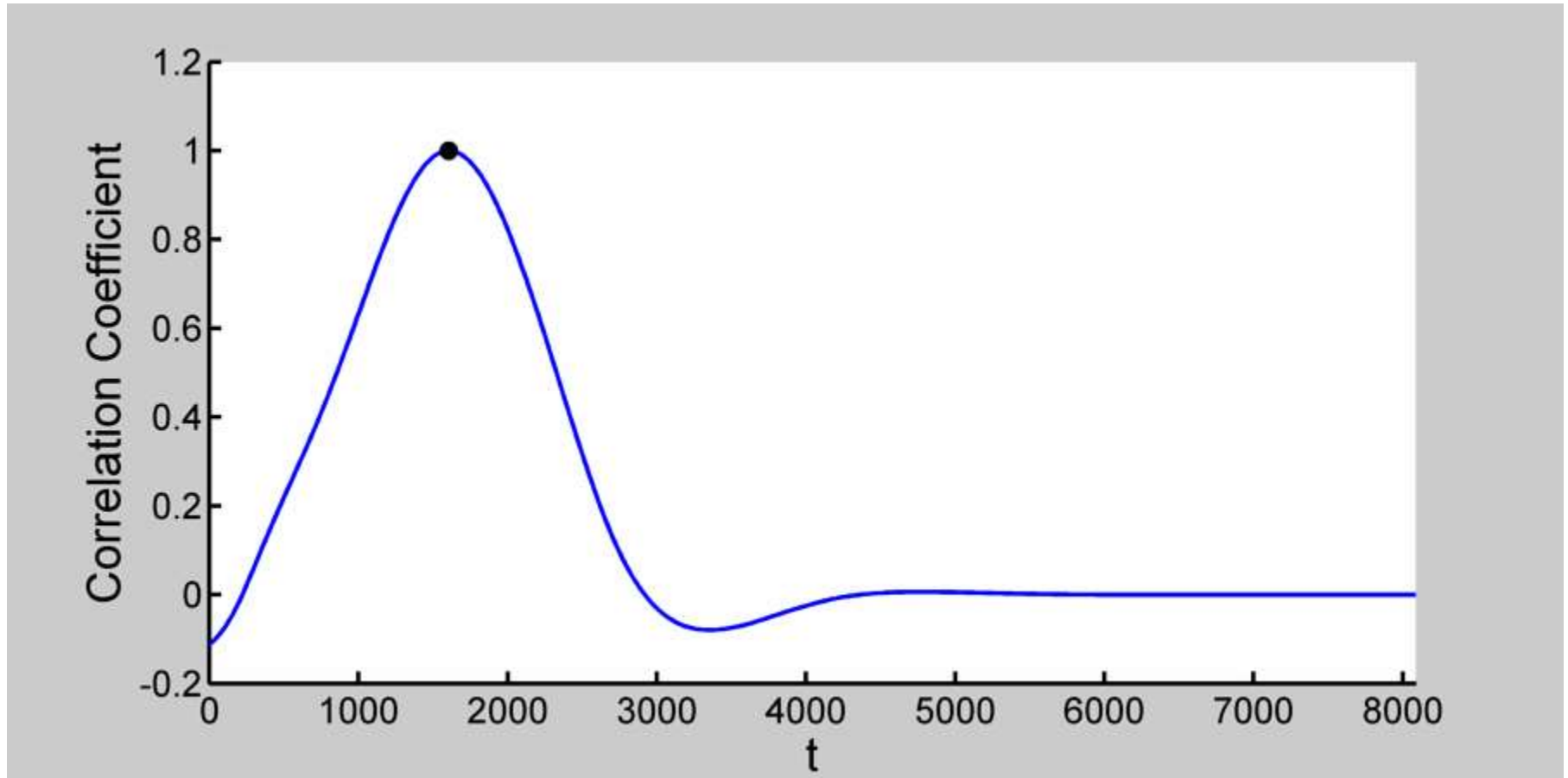
For a spline:



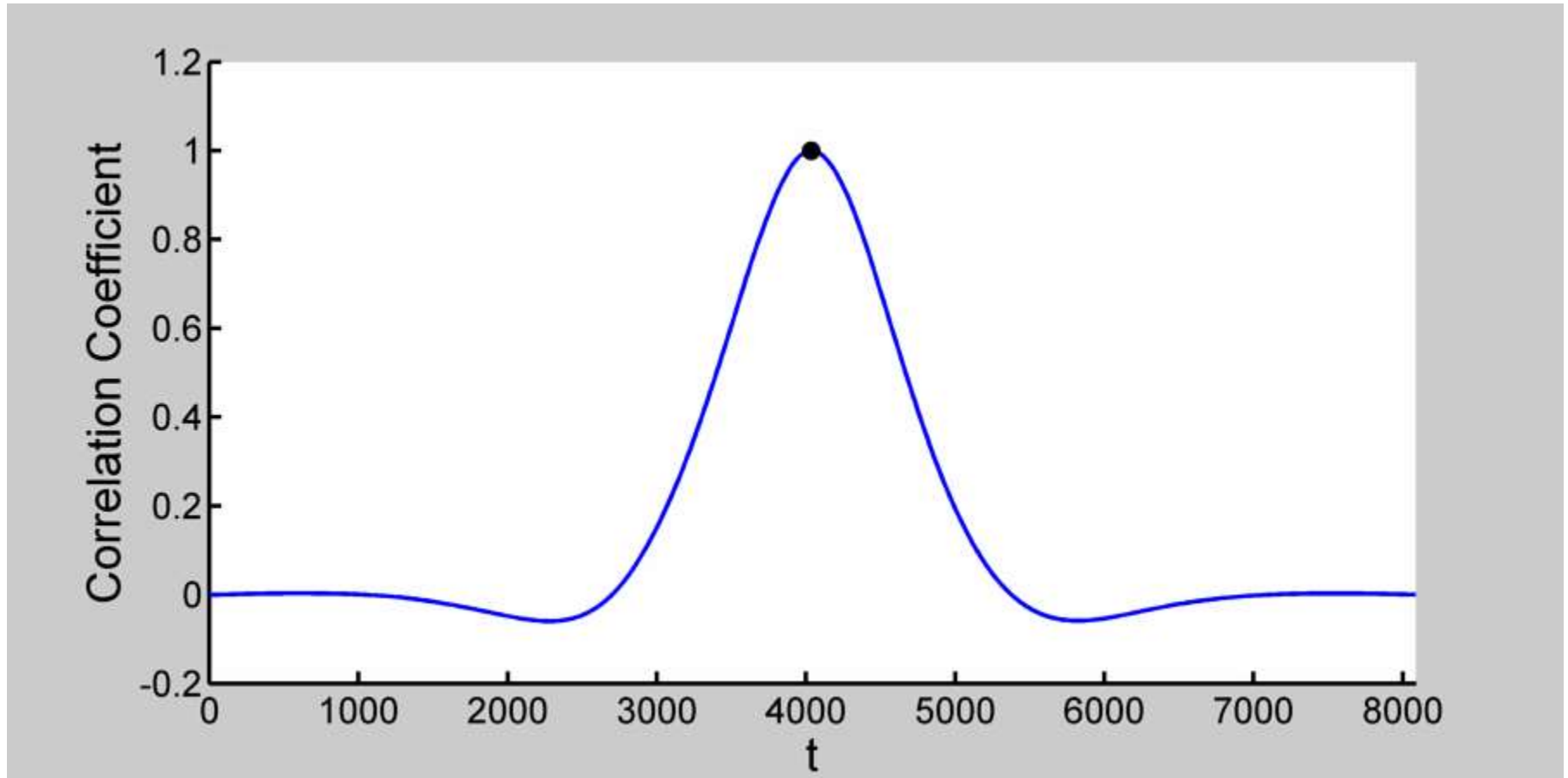
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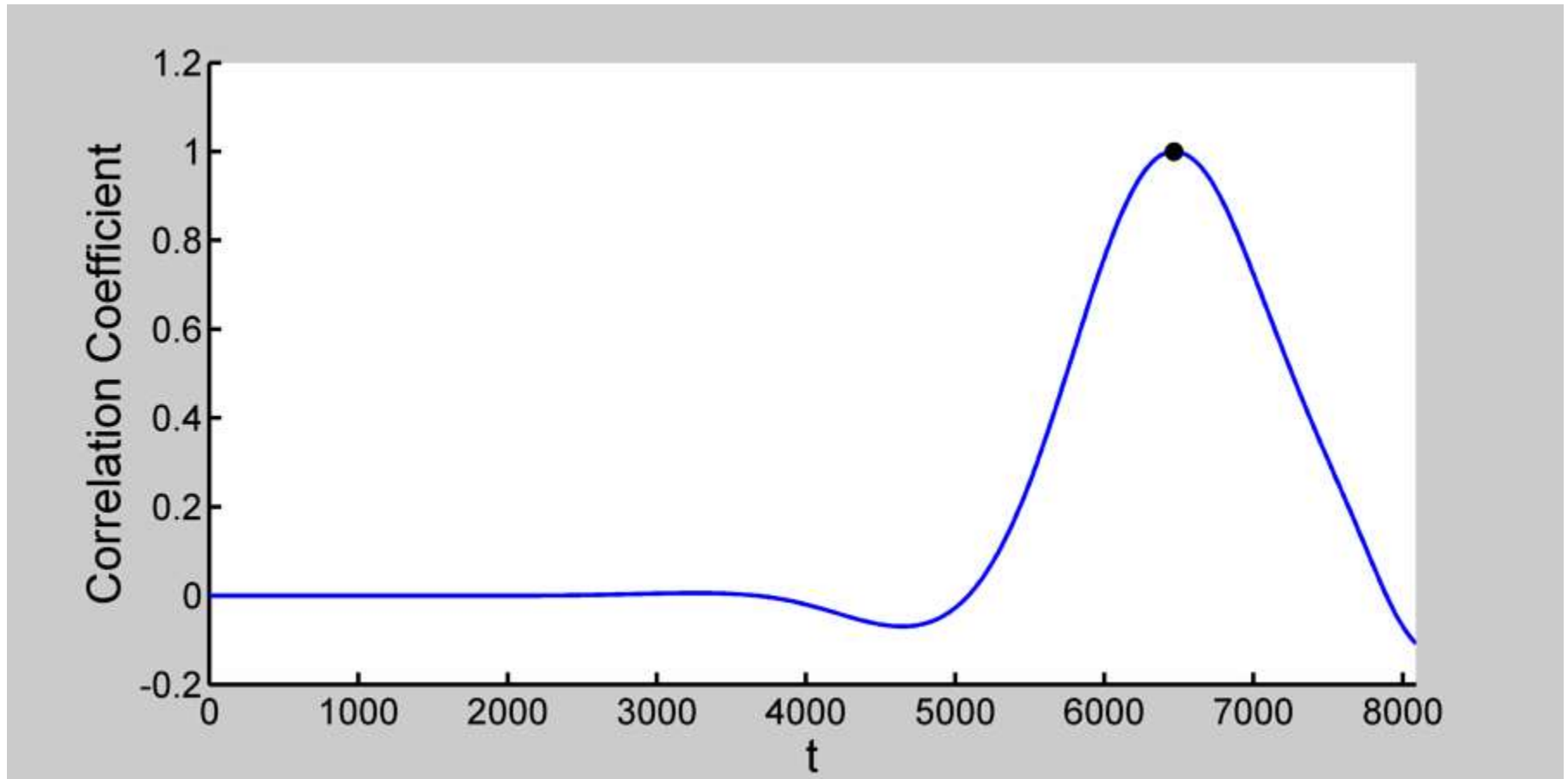
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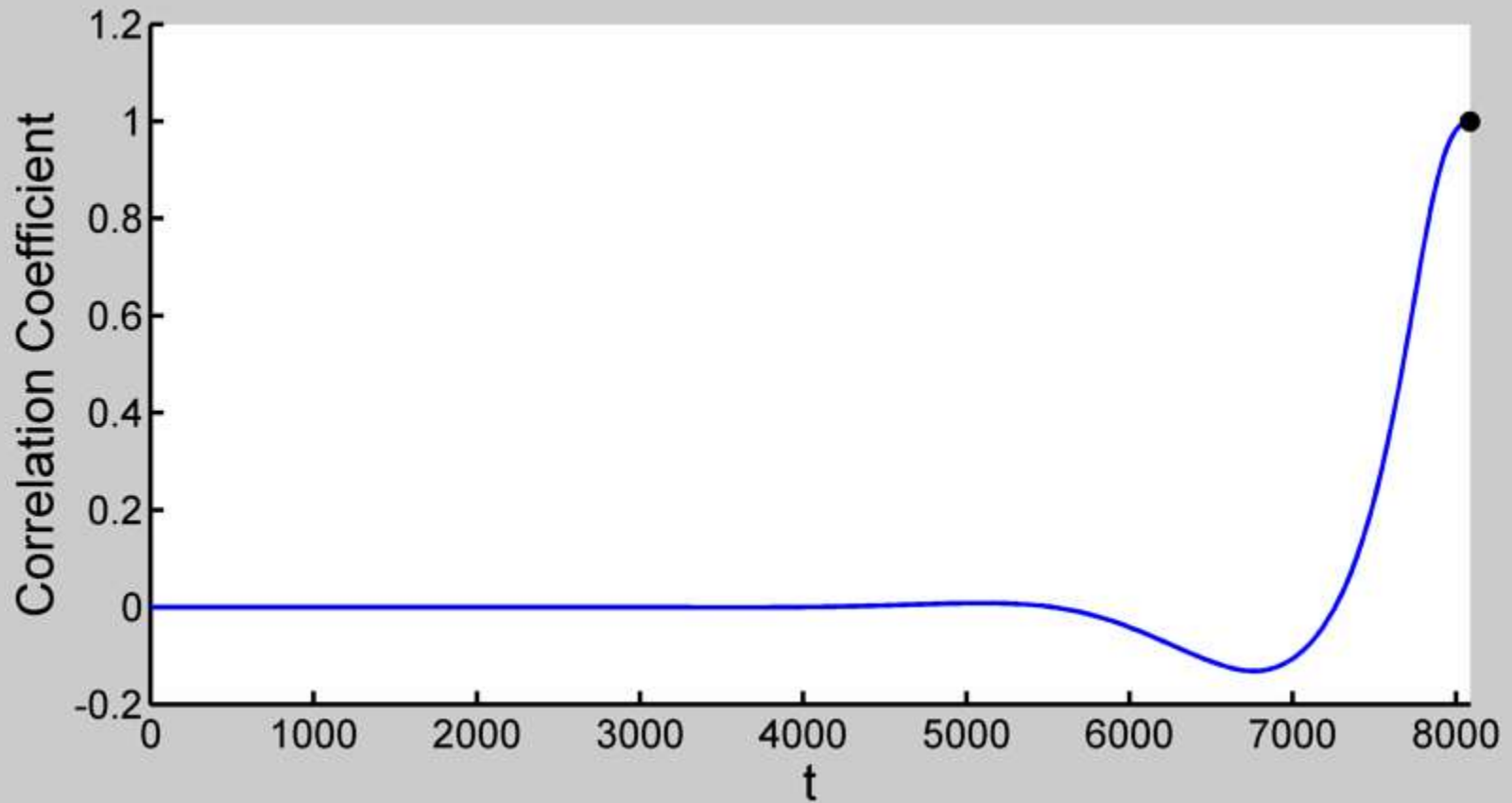
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Conclusions

- Systematic uncertainties are uncertainty correlations
- Generalized weighted means have the ability to propagate random and systematic uncertainties correctly
- Sample-standard bracketing creates correlation between unknowns
- The degree of correlation depends on the fit