

Planning and carrying out automated structure determination using SAD phasing

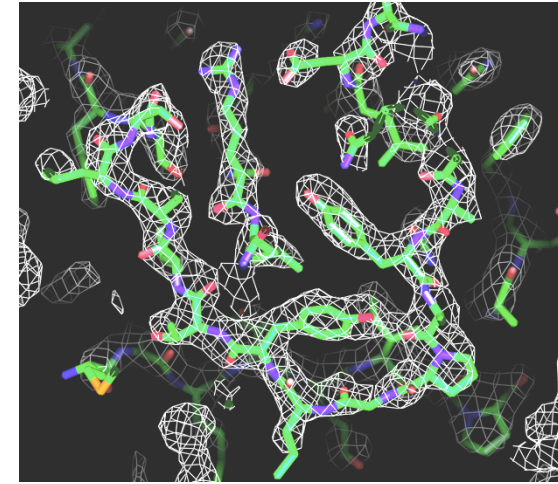
Methods and techniques in integrated structural biology:
Beyond black boxes
Strasbourg, October 5, 2016

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Los Alamos National Laboratory



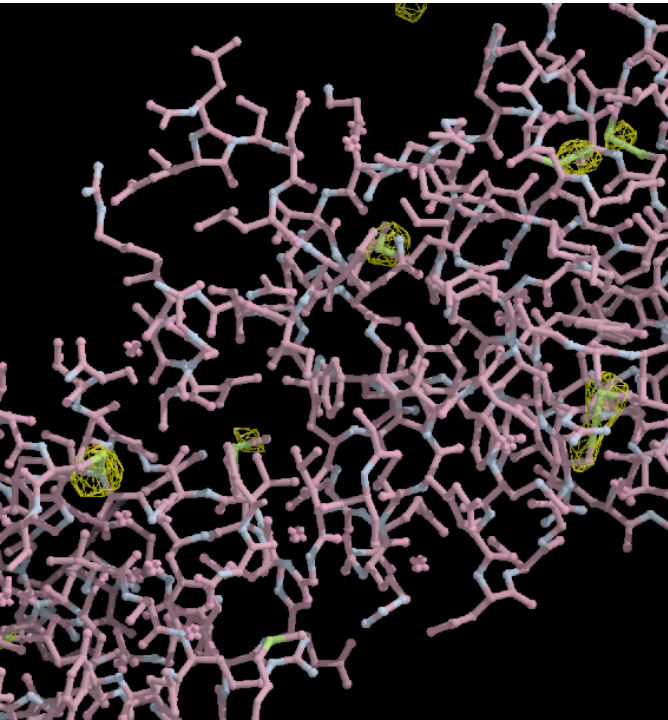
Steps in Single Wavelength Anomalous Diffraction (SAD) Structure Determination

- **Plan the experiment**
- **Measure the data**
- **Scale the data**
- **Evaluate the accuracy of the anomalous differences**
- **Find the anomalous sub-structure**
- **Identify hand of sub-structure**
- **Calculate experimental phases and a map**
- **Improve the map with density modification**
- **Build and refine a model**



Planning a SAD experiment

Will I find the sites of anomalously-scattering atoms?



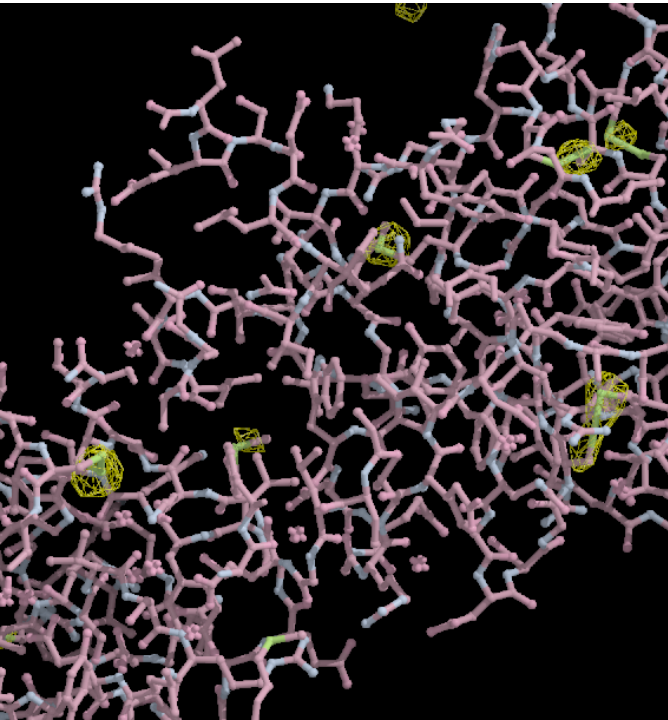
Planning a SAD experiment

How many sites?

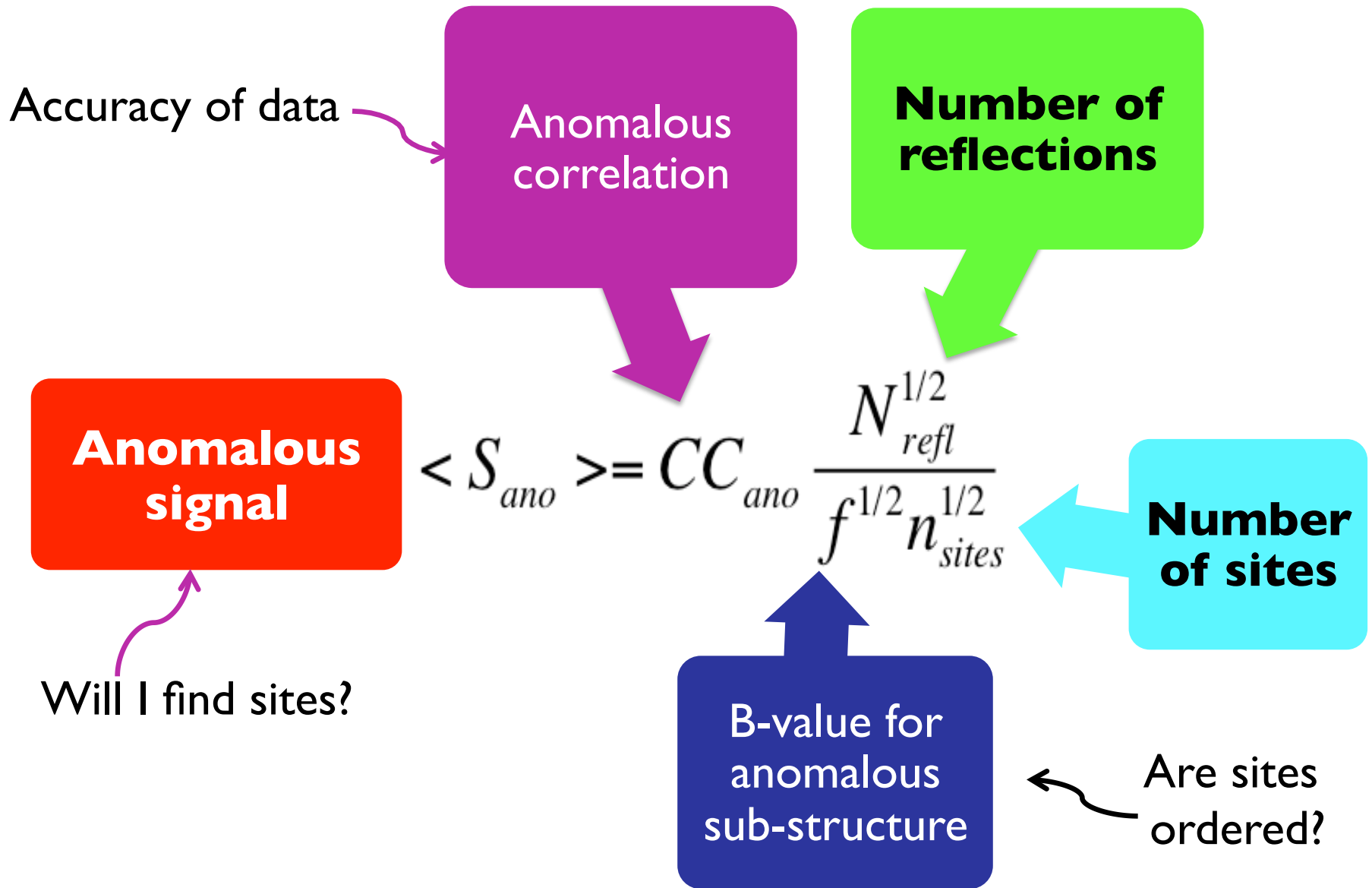
How many reflections?

Are the sites (on average) well ordered?

Are the data well-measured?



What determines if I will find the sites?



Maximizing the anomalous signal and the anomalous correlation

The **anomalous correlation** CC_{ano} is a measure of the accuracy of each anomalous difference

The **anomalous signal** is a measure of how much total information per site is present in the anomalous differences

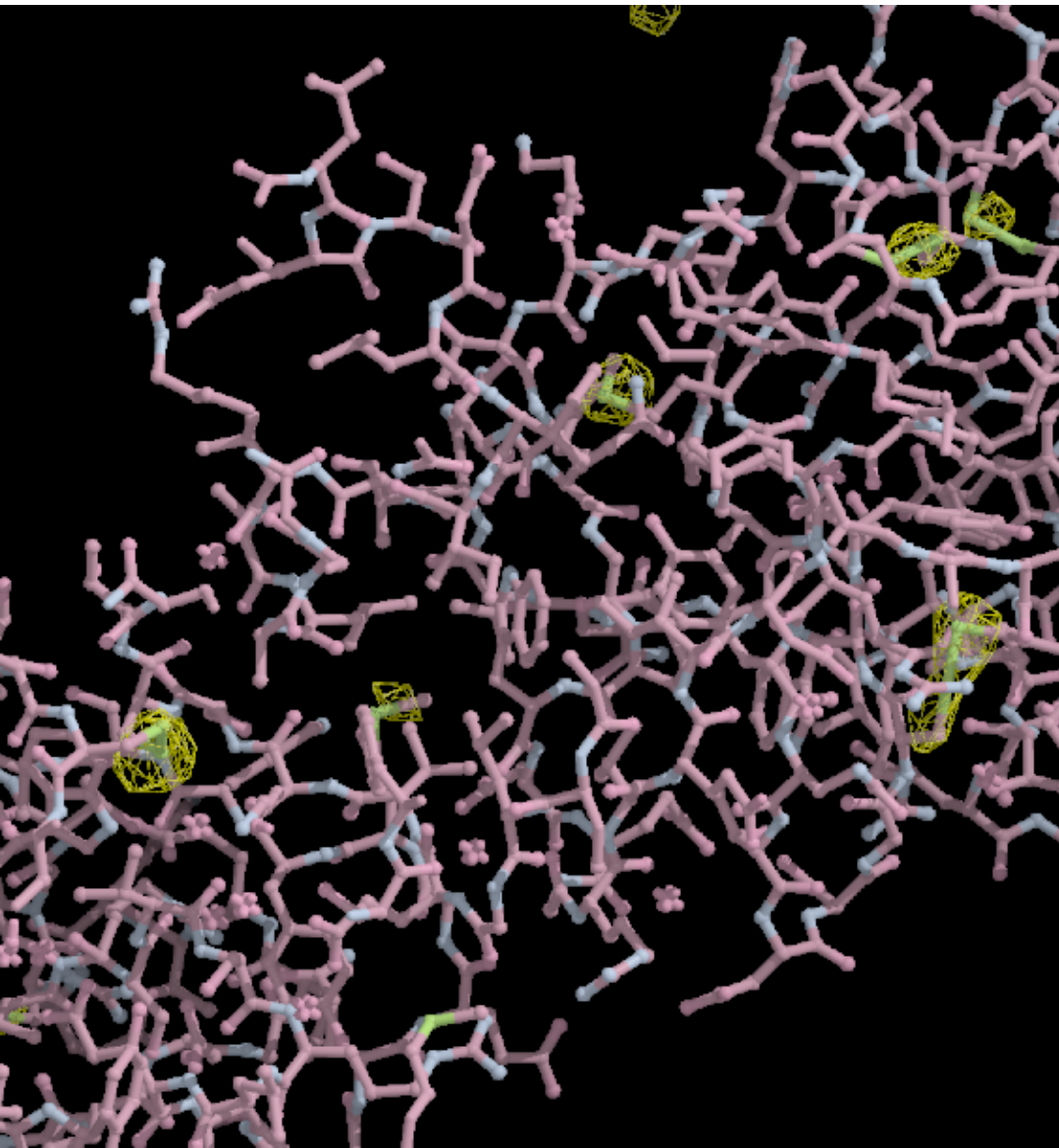
Anomalous correlation: accuracy of anomalous differences

Correlation of observed and **sub-structure** anomalous differences

$$CC_{ano} \equiv \frac{\langle \Delta_{ano,j} \Delta_{ano,j}^{obs} \rangle}{\langle \Delta_{ano} \rangle^{1/2} \langle \Delta_{ano}^{2,obs} \rangle^{1/2}}$$

CC_{ano} indicates how much of each anomalous difference is useful (on average)

Anomalous signal: peak height at coordinates of anomalously-scattering atoms



$$S_{ano} = \frac{\langle \rho_{ano}(x_j) \rangle}{\langle \rho_{ano}^2 \rangle^{1/2}}$$

Typical values of S_{ano} for solved datasets: 10-20

Anomalous difference Fourier with observed data and model phases

How big will my anomalous signal be?

Expected value of
anomalous signal S_{ano}

$$\langle S_{ano} \rangle = CC_{ano} \frac{N_{refl}^{1/2}}{f^{1/2} n_{sites}^{1/2}}$$

f is 2nd moment of the
anomalous scattering factor
(f is large if B -value for anomalously-
scattering atoms is high)

$$f = \frac{\langle (f^h)^2 \rangle}{\langle f^h \rangle^2}$$

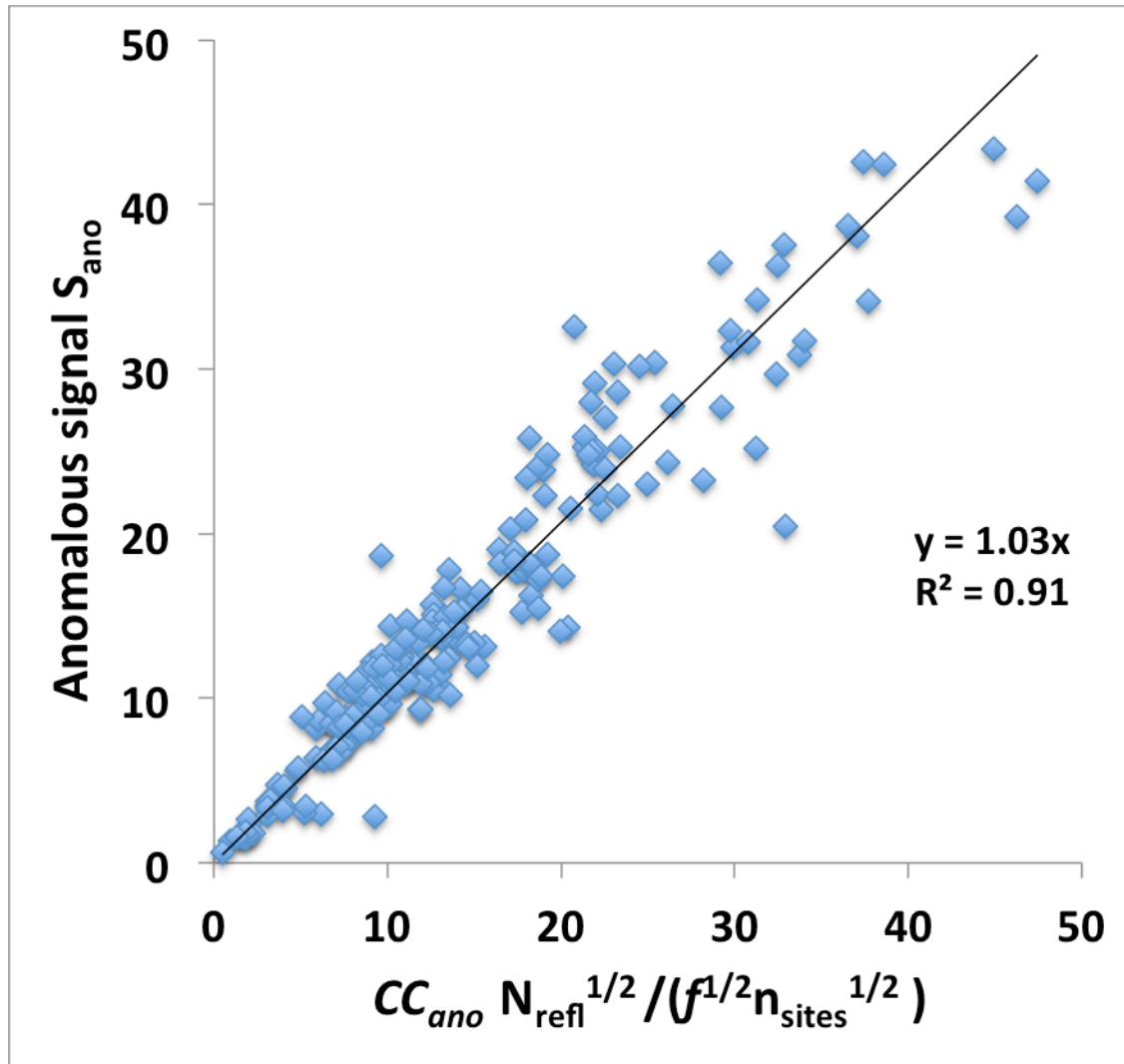
(f^h in this equation is the
anomalous scattering factor)

$$f^h \equiv f'' e^{-B (\sin^2 \theta_h / \lambda^2)}$$

Perfect data (20,000 reflections, 8 sites): $S_{ano} = (20000/8)^{1/2} = 50$

Good data (overall $CC_{ano} = 0.36$ $f = 2.0$): $S_{ano} = 12.6$

Checking our simple model for anomalous signal



$$\langle S_{ano} \rangle = CC_{ano} \frac{N_{refl}^{1/2}}{f^{1/2} n_{sites}^{1/2}}$$

CC_{ano} : Correlation of anomalous differences with model differences

S_{ano} : Peak height in model-phased difference Fourier

218 SAD datasets 1.2 – 4.5 Å

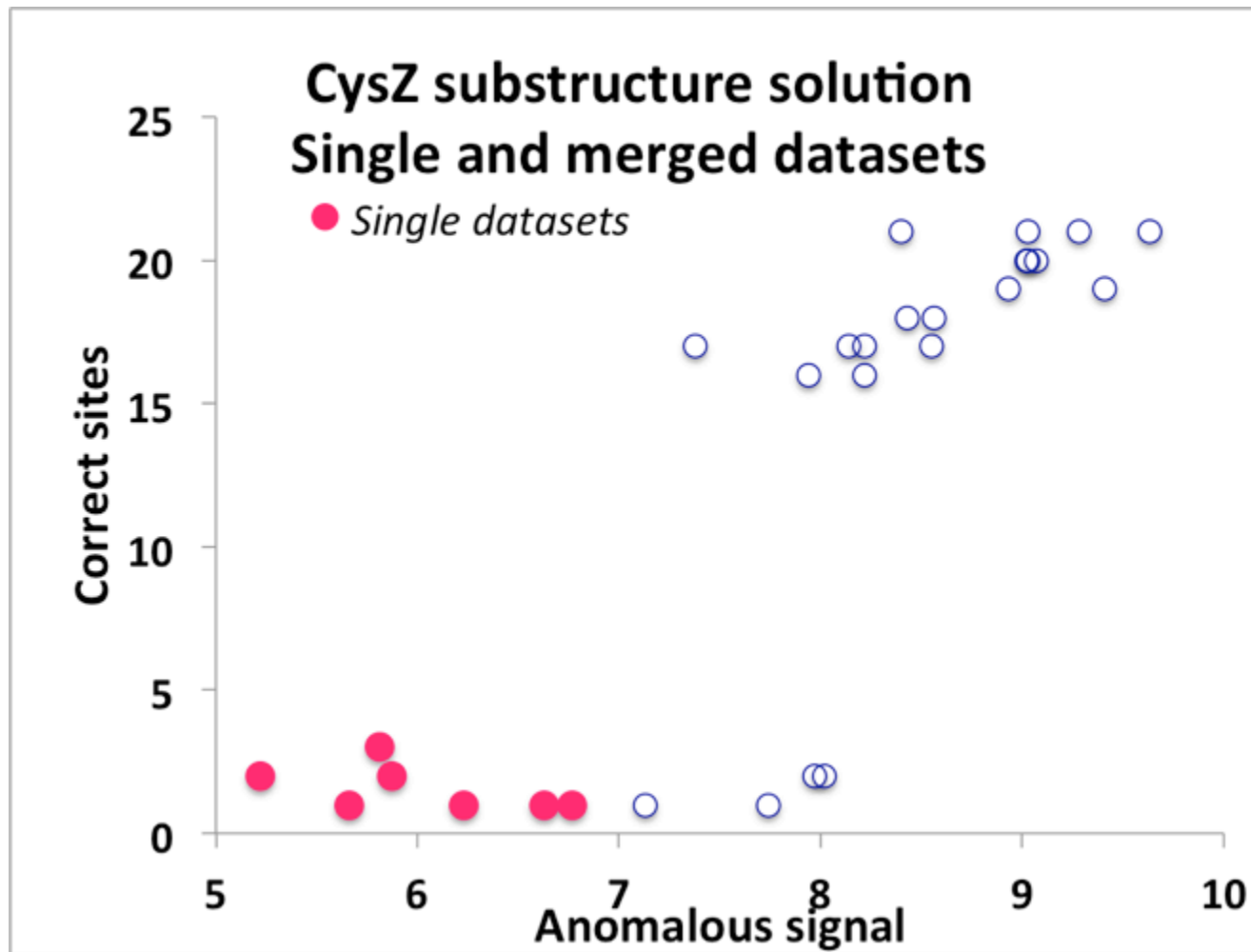
CysZ multi-crystal sulfur-SAD data

Qun Liu, Tassadite Dahmane, Zhen Zhang, Zahra Assur, Julia Brasch, Lawrence Shapiro, Filippo Mancini, Wayne Hendrickson (2012). Science 336, 1033-1037

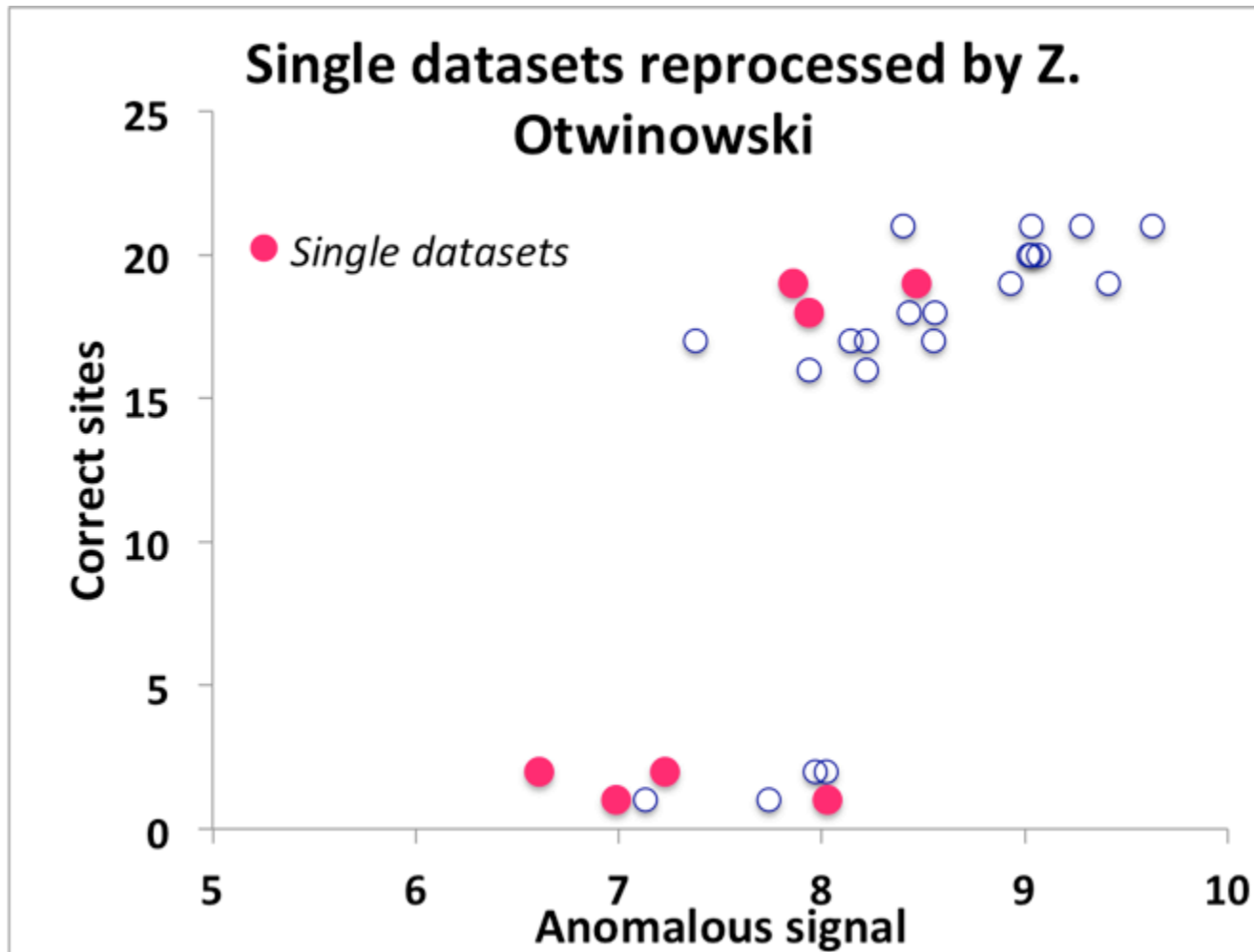
Data from 7 crystals collected at wavelength of 1.74 Å to resolution of 2.3 Å

Can anomalous signal tell us which merged datasets will be solved?

CysZ multi-crystal sulfur-SAD data

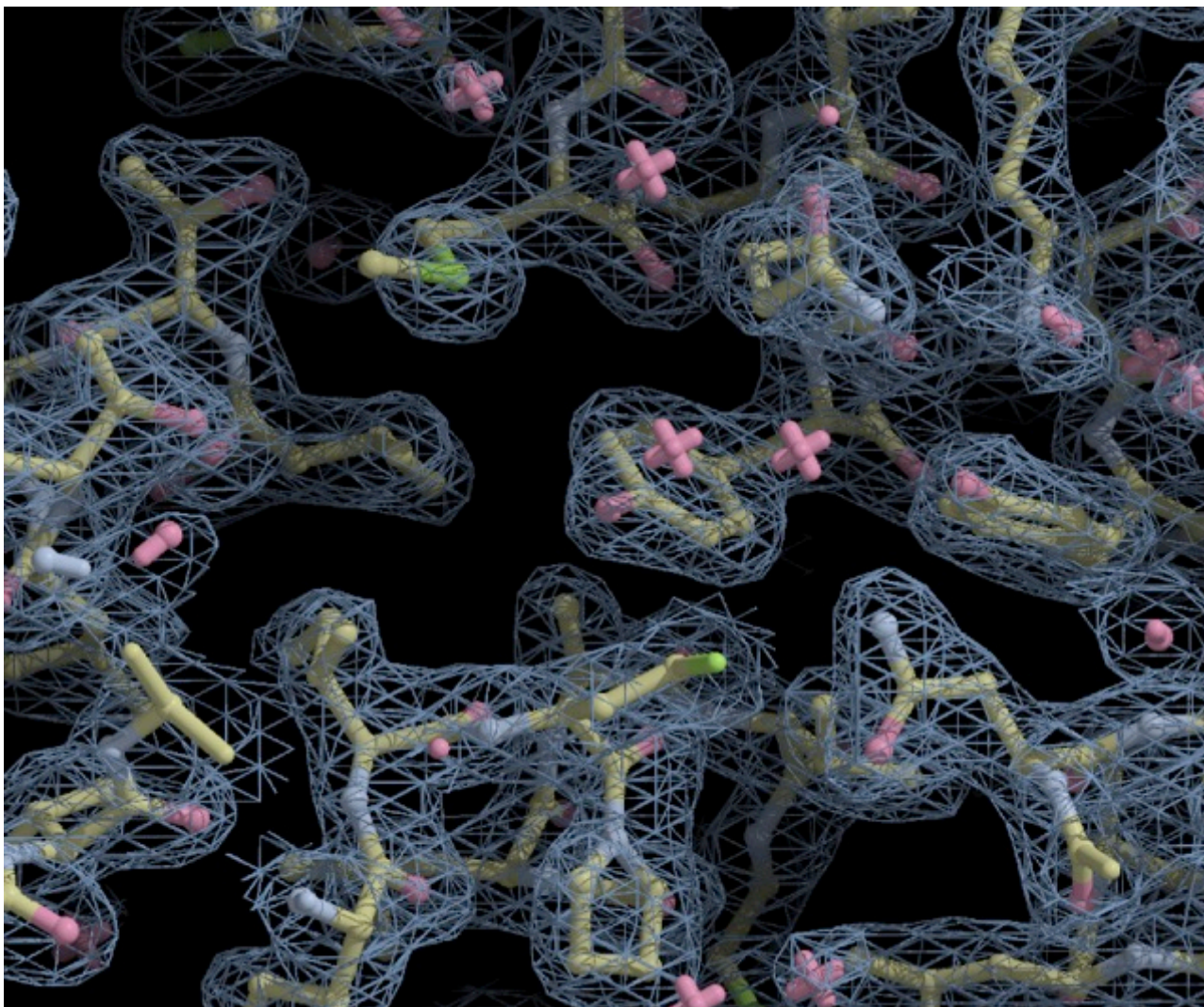


CysZ multi-crystal sulfur-SAD data



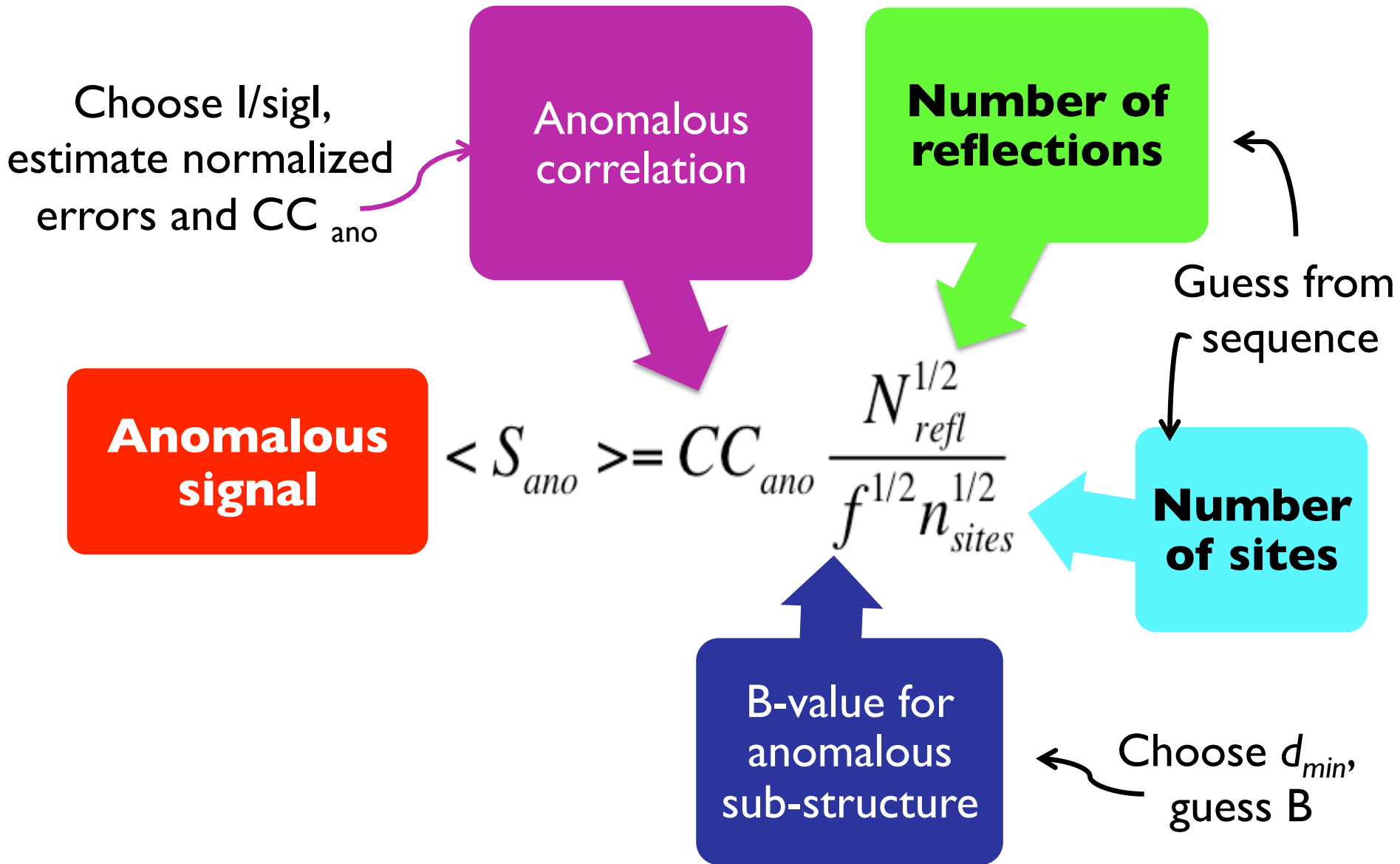
CysZ single-crystal sulfur-SAD data

Crystal 6 *AutoSol R/Rfree=0.24/0.27*



phenix.plan_sad_experiment

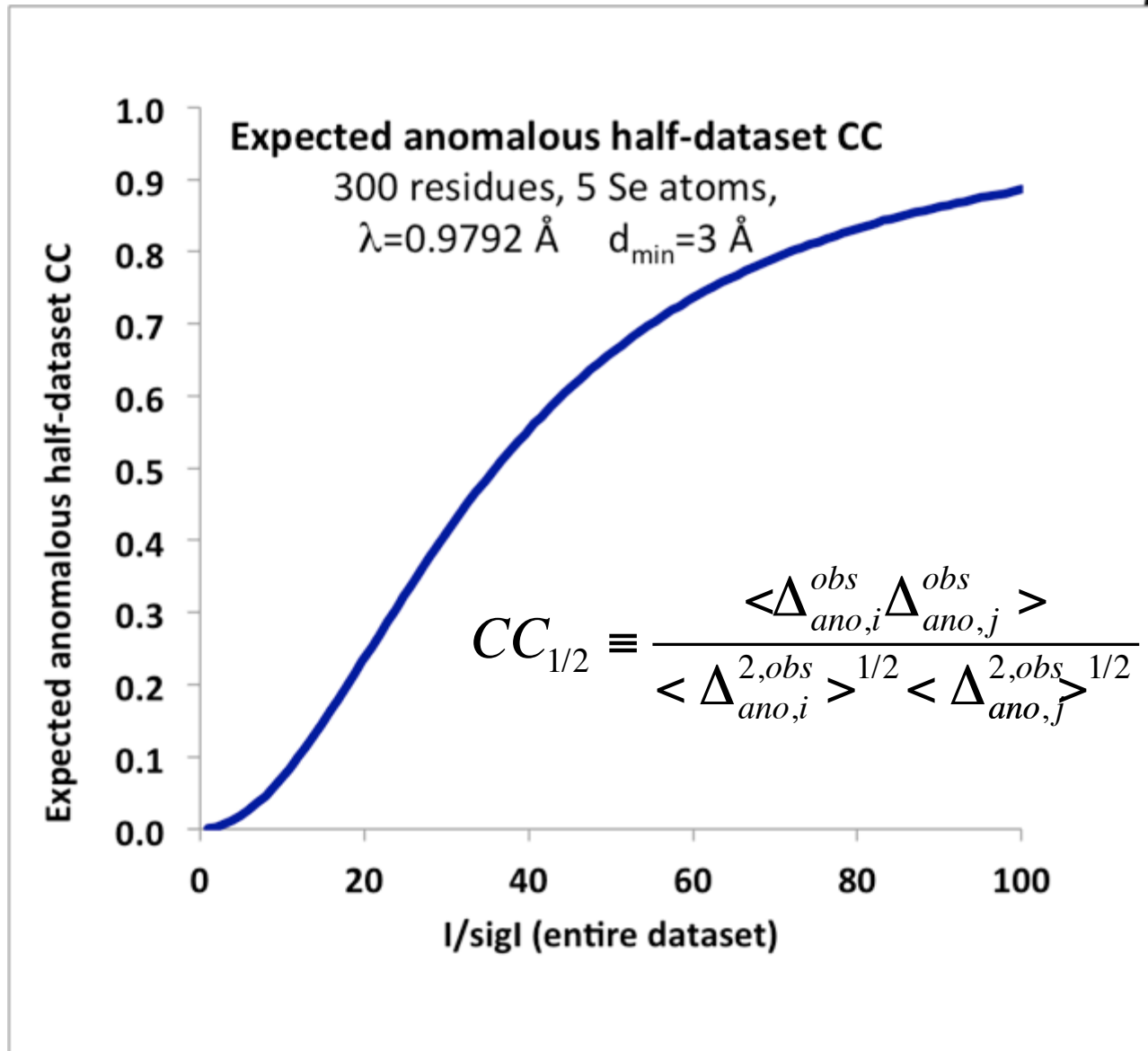
Design an experiment that will give you enough anomalous signal



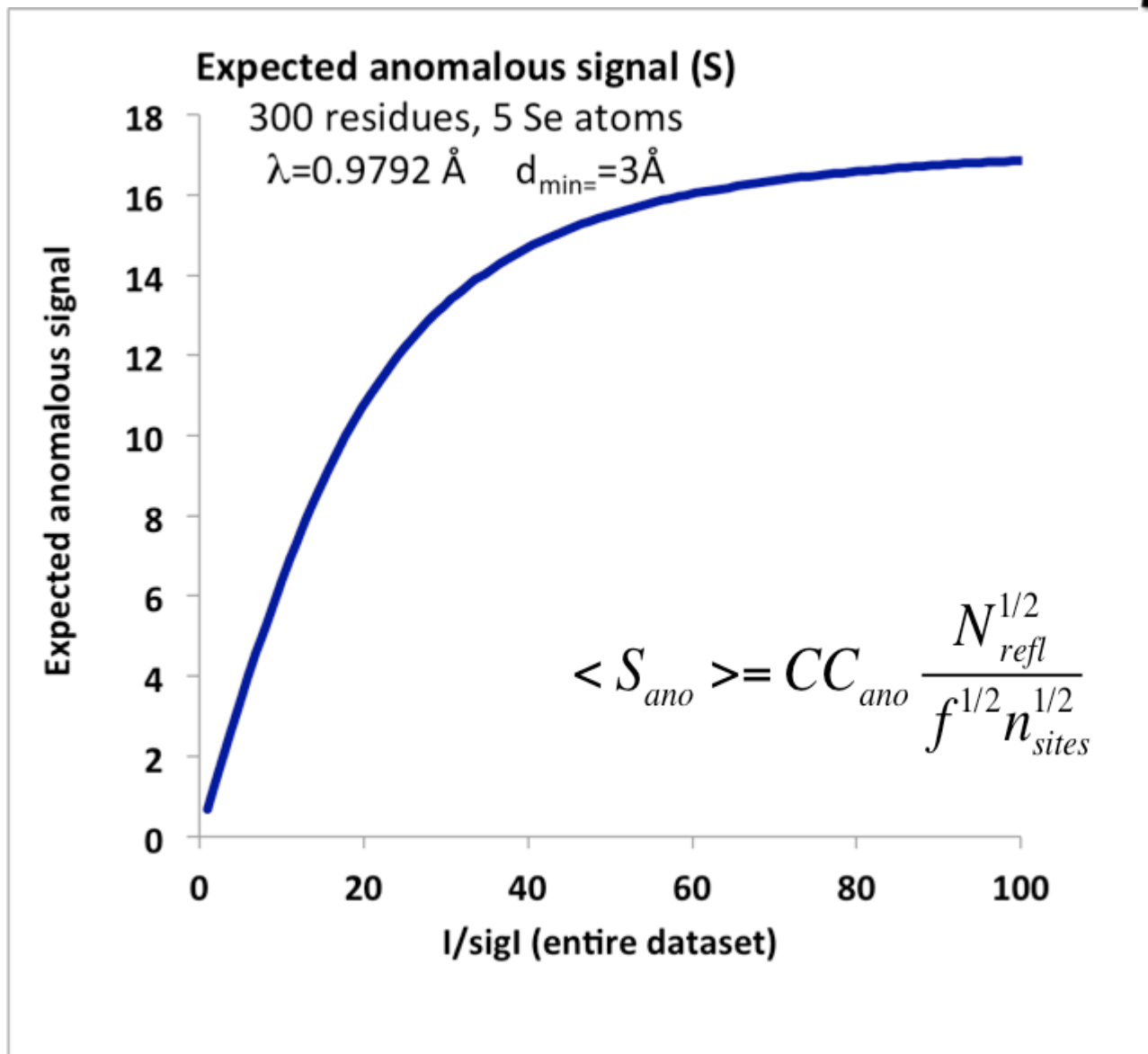
Will I solve my structure?

Simulate experiment with
phenix.plan_sad_experiment based on:

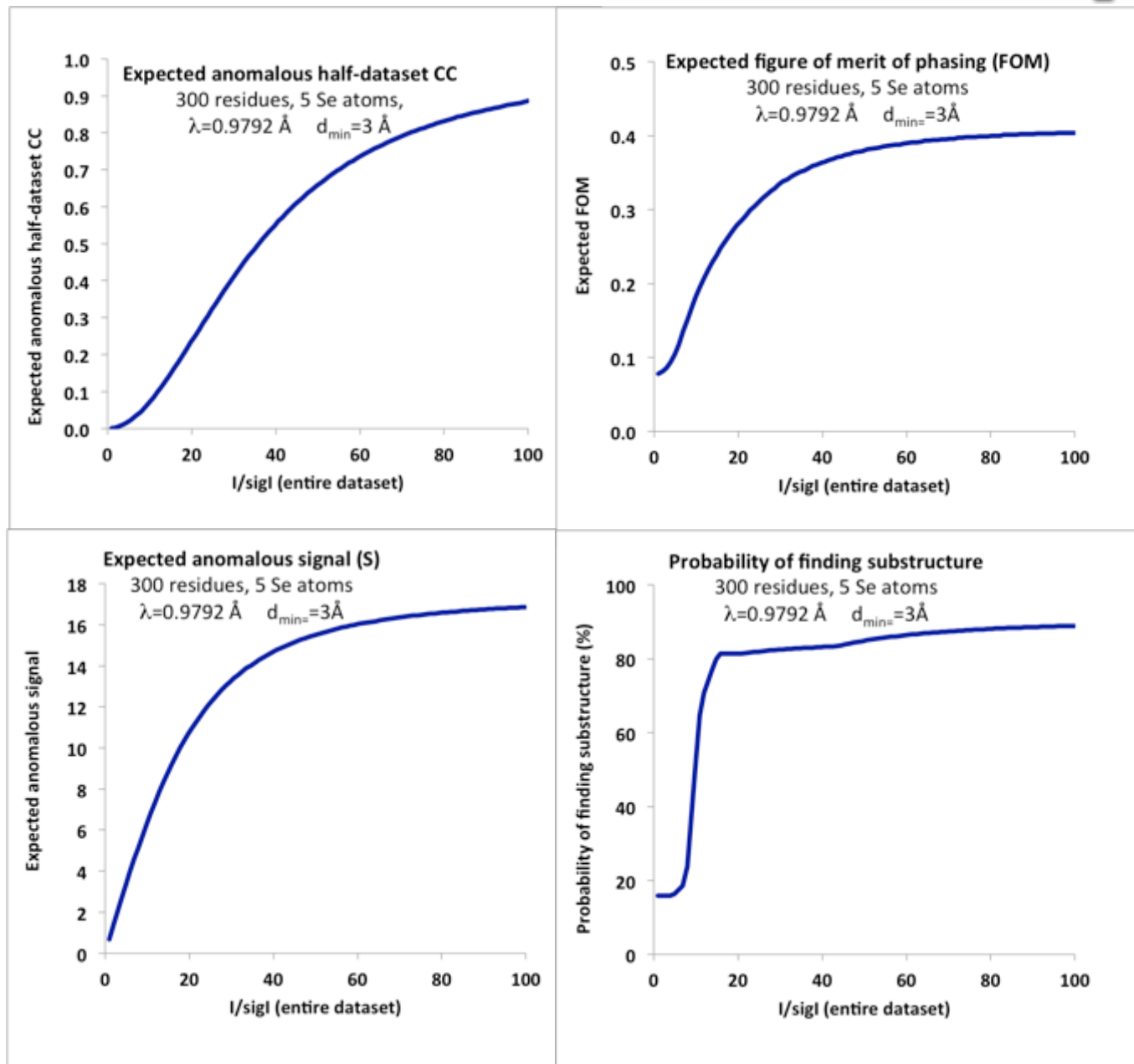
- I/σ (errors in measurement)
- Anomalous-scattering atom (f'')
- Sequence (other atoms)
- Resolution of data
- Number of sites



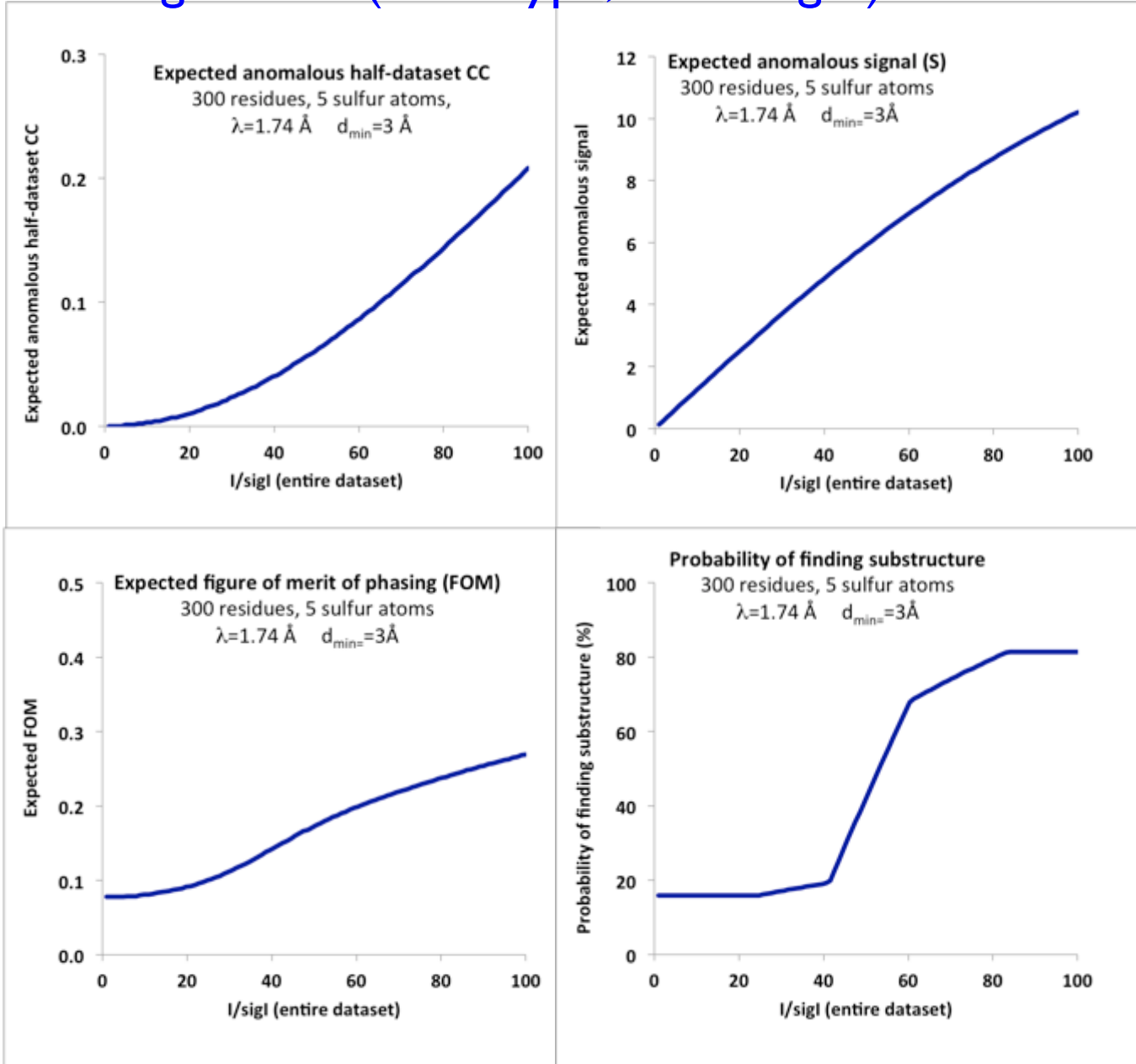
Anomalous signal depends on I/sigI



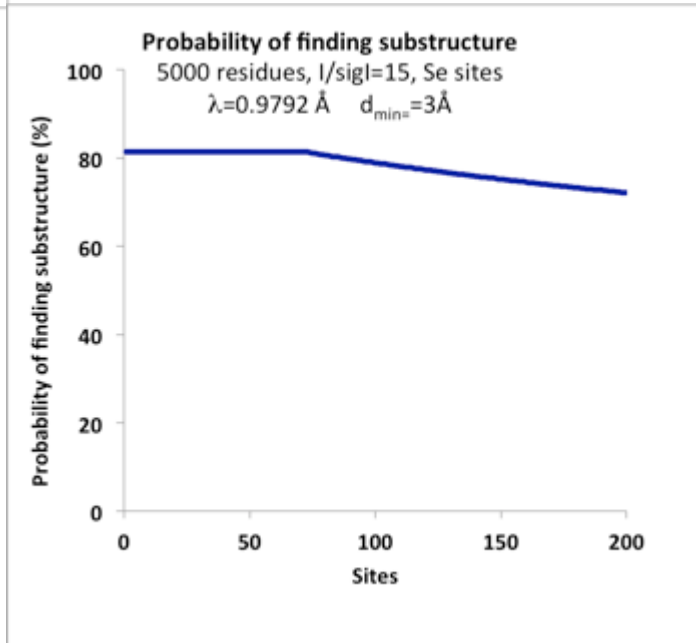
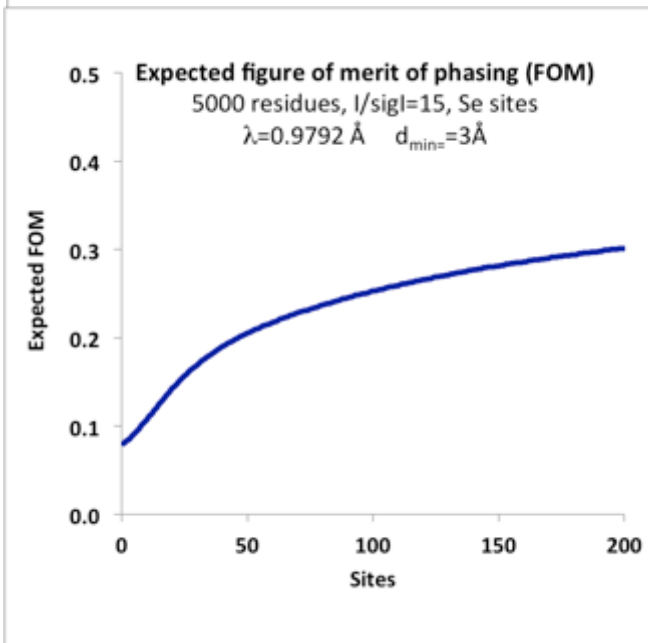
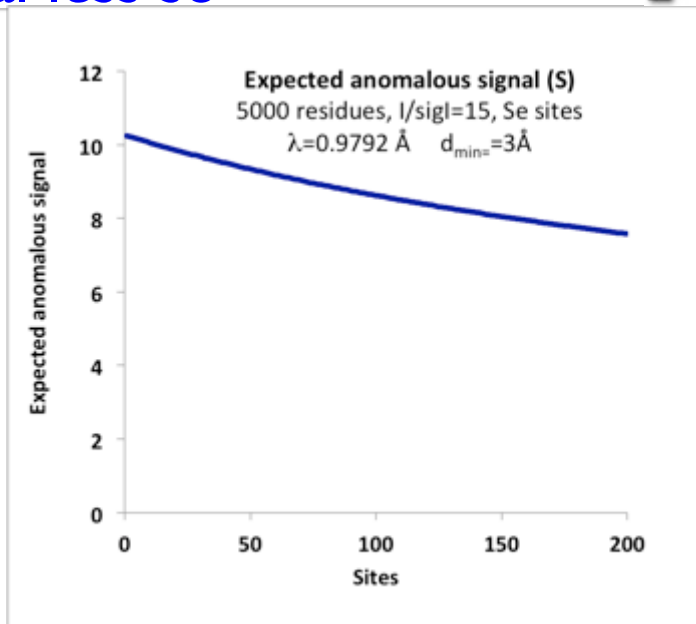
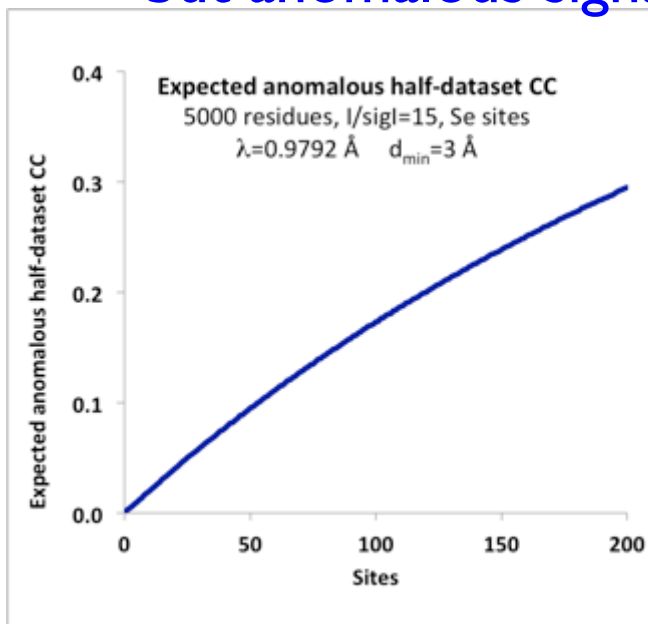
Probability of finding substructure depends on I/sigI



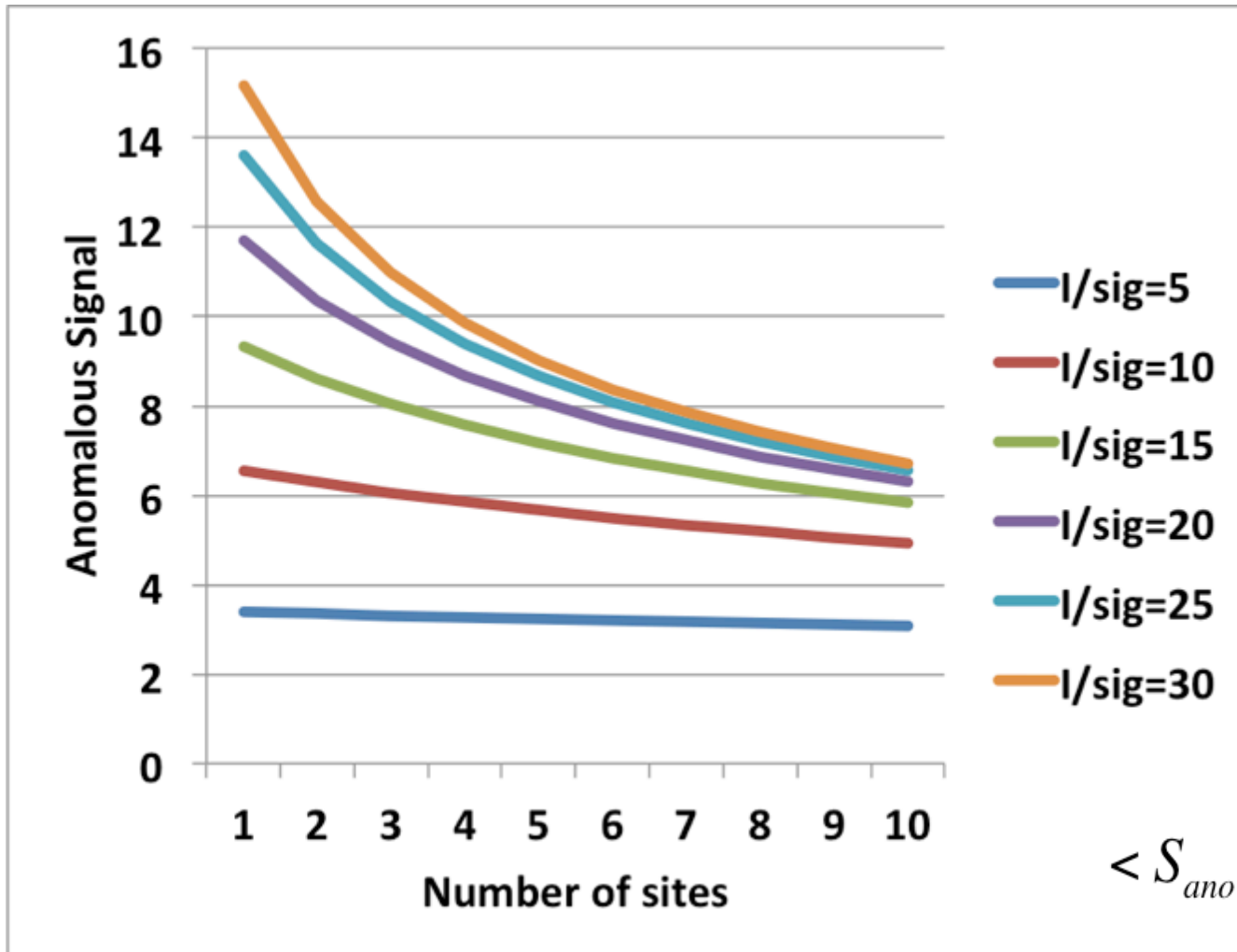
Anomalous data quality depends on I/σ_I ... and scattering factors (atom type, wavelength)



Phasing quality depends a lot on number of sites... but anomalous signal less so

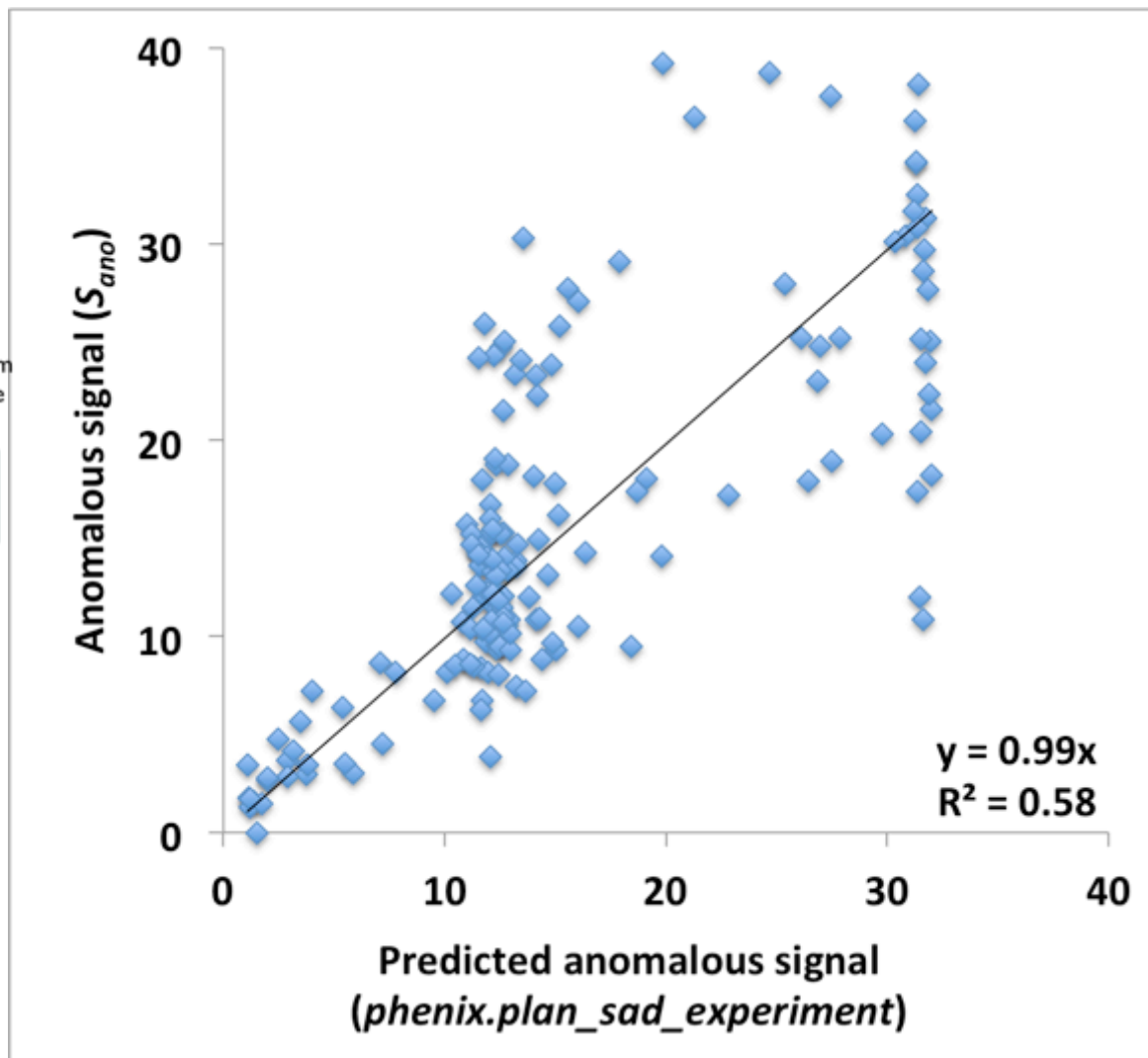
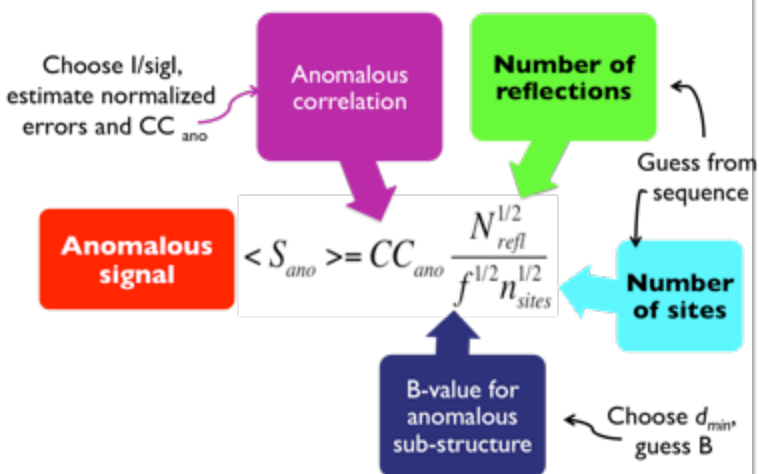


Anomalous signal vs I/σ and sites
 100 residues, varying S_e , varying I/σ

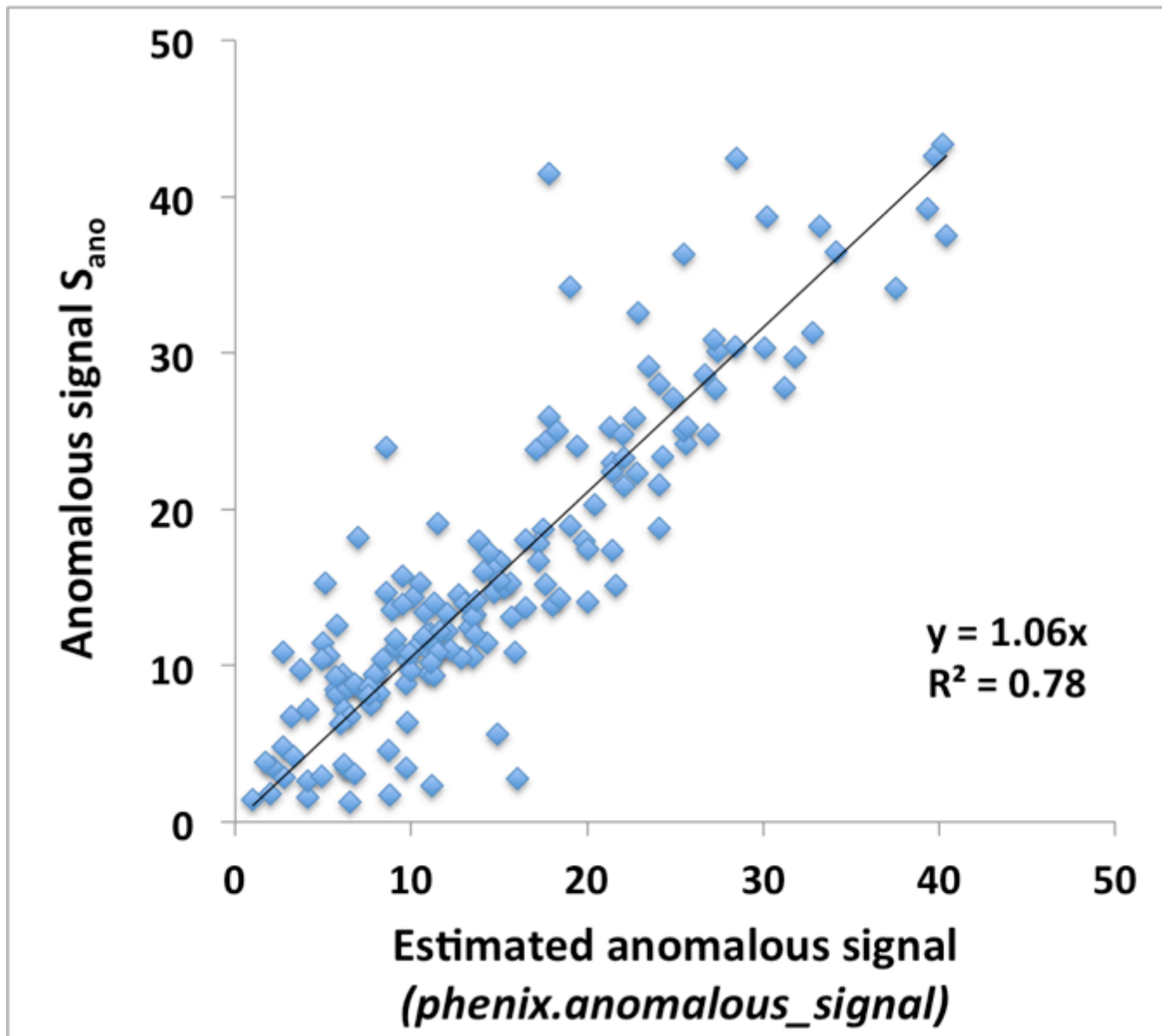


$$\langle S_{ano} \rangle = CC_{ano} \frac{N_{refl}^{1/2}}{f^{1/2} n_{sites}^{1/2}}$$

Estimating the anomalous signal before collecting the data



Estimating the anomalous signal after collecting the data



Finding the anomalous sub-structure

Using the SAD likelihood function to find sites

***“The likelihood of measuring the observed
anomalous data***

given

a potential sub-structure”

Using the SAD likelihood function to find the anomalous sub-structure

Start with guess about the anomalous sub-structure

From anomalous difference Patterson

Random

Any other source

Find additional sites that increase the likelihood

*LLG completion based on log-likelihood gradient maps**

Iterative addition of sites

Related to using an anomalous difference Fourier—but better

*La Fortelle, E. de & Bricogne, G. (1997). *Methods Enzymol.* 276, 472-494
McCoy, A. J. & Read, R. J. (2010). *Acta Cryst.* D66, 458-469.

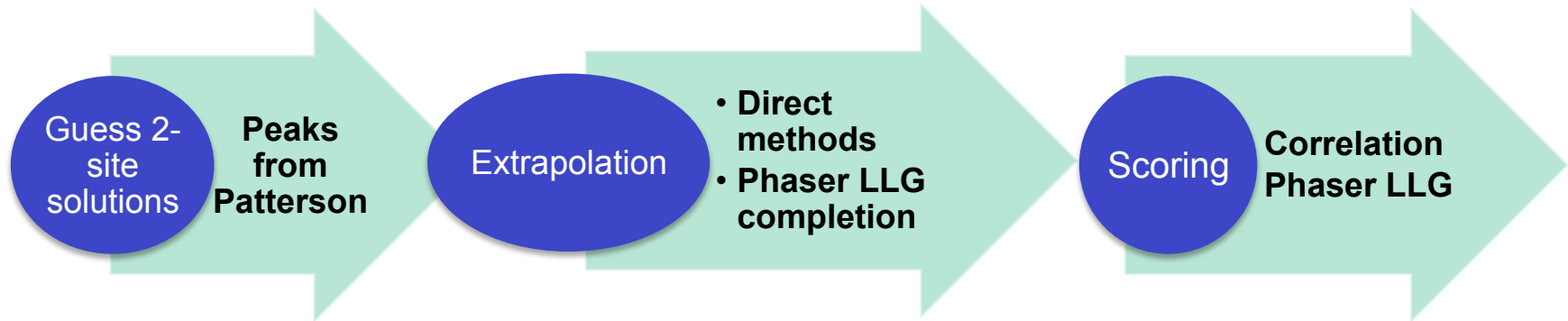
LLG sub-structure searches in HySS

Test cases

164 SAD datasets from PDB (largely JCSG MAD data)

Using peak, remotes, inflection as available to include data
with low anomalous signal

Finding anomalous substructure with LLG completion



- **Range of resolution**
Variable number of Patterson solutions

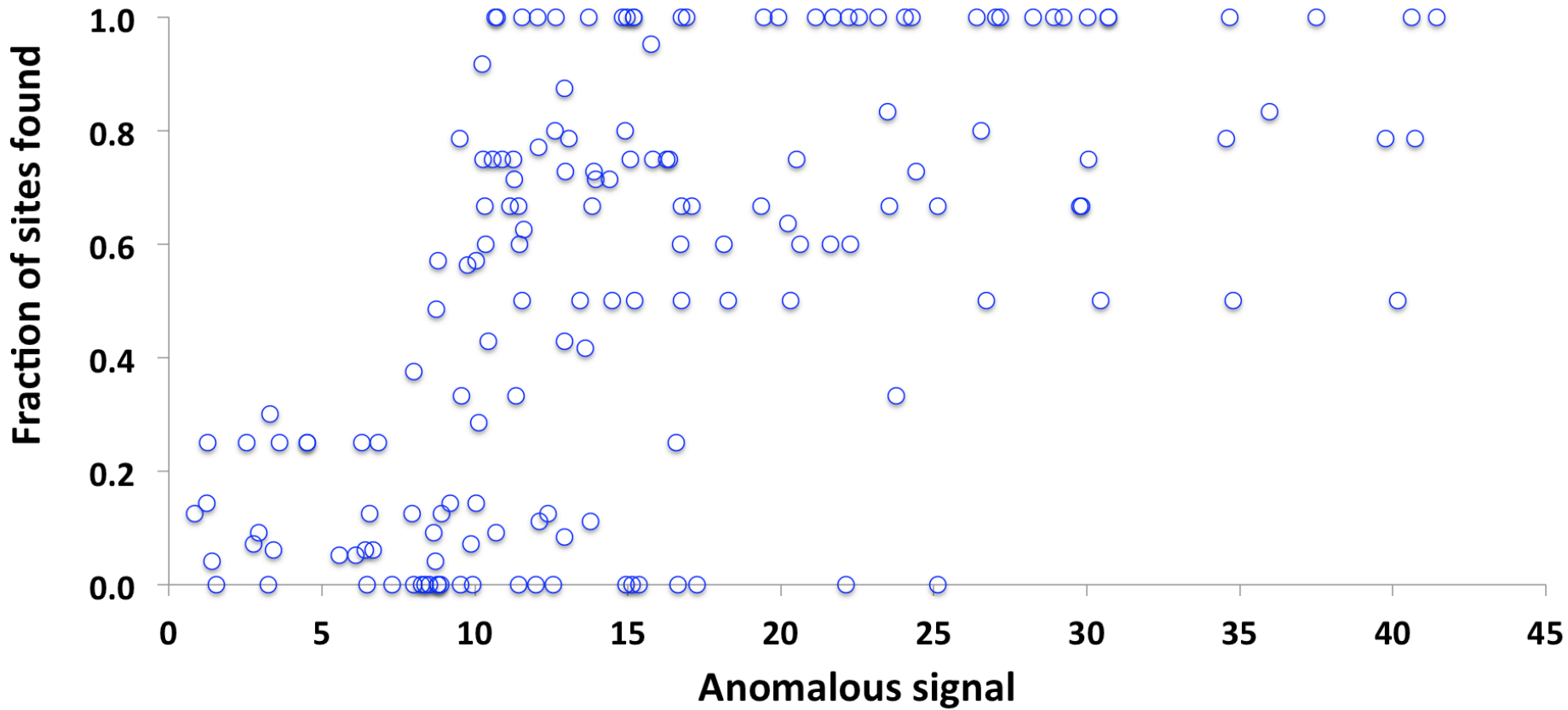
**Adjustable
LLGC_SIGMA
(cut-off for peak height)**

**Use LLG score to
compare solutions**

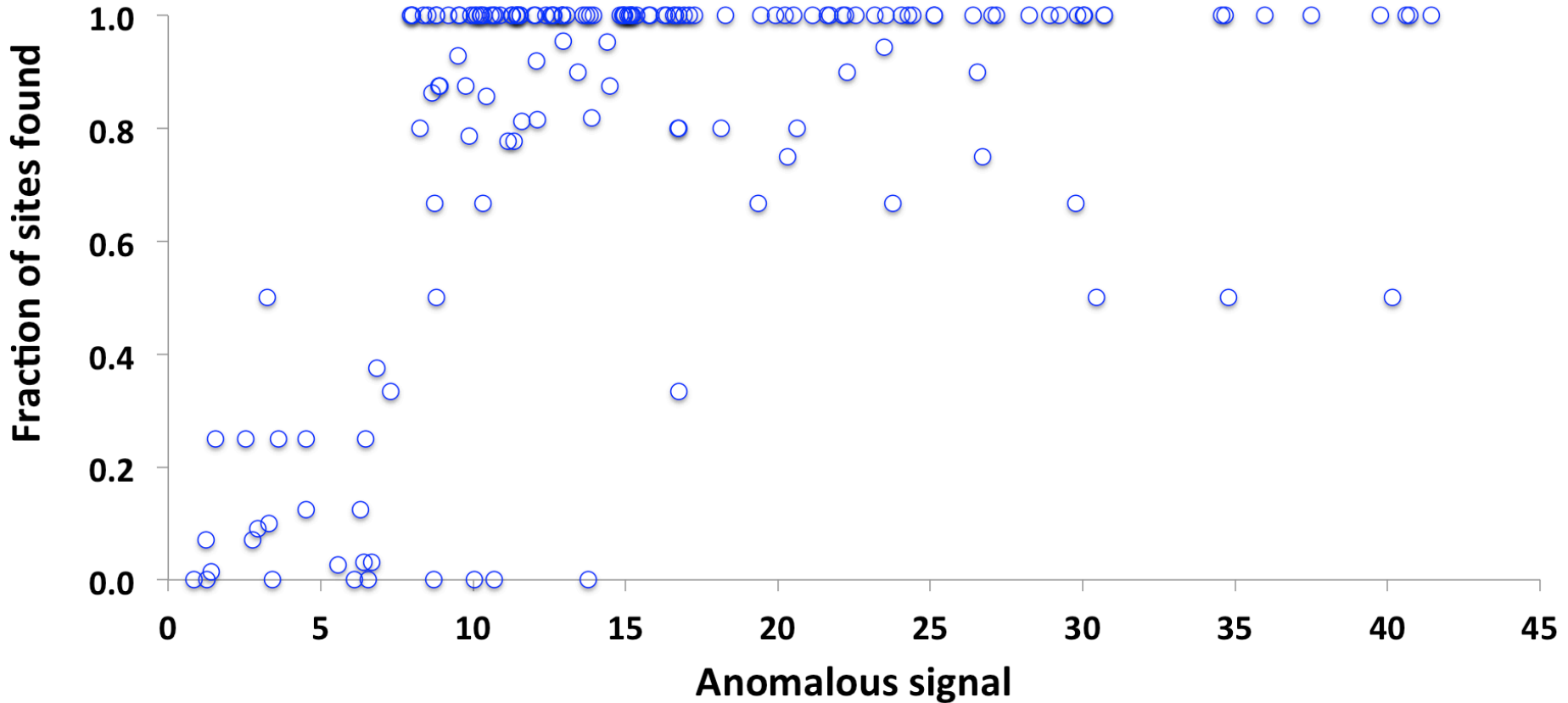
**Terminate early if same
solution found several
times**

**Run quick direct
methods first**

Dual Space Sub-structure Completion

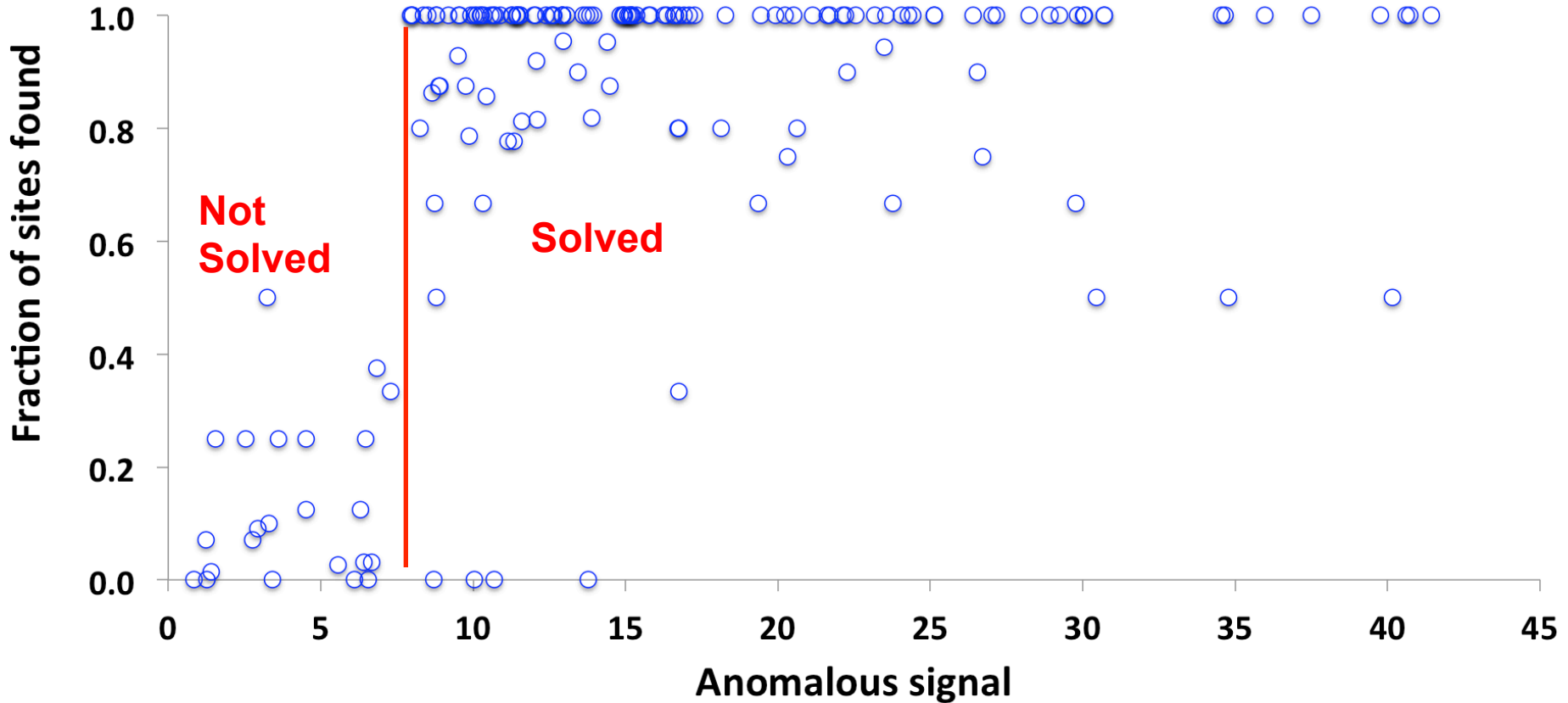


LLG Sub-structure Search



Bunkóczy et al., Nature Methods 12, 127–130 (2015).

Anomalous signal indicates if a dataset can be solved



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