



From SEDFIT to SEDPHAT

SEDPHAT for AUC SV Data Analysis for Equilibrium Systems

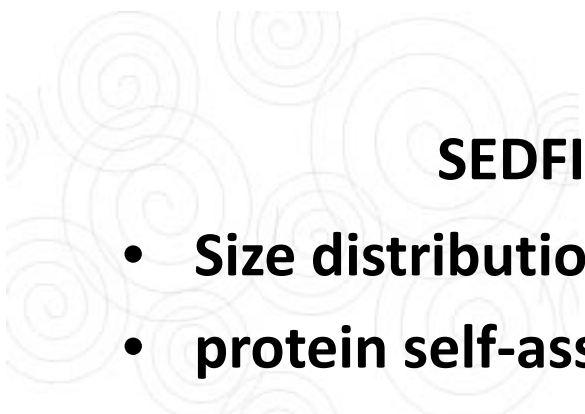
<http://www.analyticalultracentrifugation.com>

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SEDFIT

- **Size distribution analysis**
- **protein self-association**
- **size-and-shape distributions**
- **macromolecules**
- **sedimentation velocity ultracentrifugation**
- **Lamm equation modeling and**

SEDPHAT

- **multi-site binary and ternary protein interactions**
- **heterogeneous protein-protein interactions**
- **multi-protein complexes by multi-signal**
- **specific models**

Papers with applications of SEDFIT

- **Uncontrolled Zinc- and Copper-Induced Oligomerisation of the Human Complement Regulator Factor H and Its Possible Implications for Function and Disease.** *Journal of Molecular Biology*, Volume 384, Issue 5, 31 December 2008, Pages 1341-1352. Ruodan Nan, Jayesh Gor, Imre Lengyel, Stephen J. Perkins
- **Specificity and Reactivity in Menaquinone Biosynthesis: The Structure of *Escherichia coli* MenD (2-Succinyl-5-Enolpyruvyl-6-Hydroxy-3-Cyclohexadiene-1-Carboxylate Synthase).** *Journal of Molecular Biology*, Volume 384, Issue 5, 31 December 2008, Pages 1353-1368. Alice Dawson, Paul K. Fyfe, William N. Hunter
- **Evidence for the Oligomeric State of 'Elastic' Titin in Muscle Sarcomeres.** *Journal of Molecular Biology*, Volume 384, Issue 2, 12 December 2008, Pages 299-312. Ahmed Houmeida, Andy Baron, Jeff Keen, G Nasir Khan, Peter J. Knight, Walter F. Stafford III, Kavitha Thirumurugan, Beatrix Thompson, Larissa Tskhovrebova, John Trinick
- **The High-Resolution NMR Structure of the Early Folding Intermediate of the *Thermus thermophilus* Ribonuclease H.** *Journal of Molecular Biology*, Volume 384, Issue 2, 12 December 2008, Pages 531-539 Zheng Zhou, Hanqiao Feng, Rodolfo Ghirlando, Yawen Bai
- **NUTS and BOLTS: Applications of fluorescence-detected sedimentation.** *Analytical Biochemistry*, *In Press, Uncorrected Proof*, Available online 6 December 2008. Rachel R. Kroe, Thomas M. Laue
- ***Helicobacter pylori* neutrophil-activating protein promotes myeloperoxidase release from human neutrophils.** *Biochemical and Biophysical Research Communications*, Volume 377, Issue 1, 5 December 2008, Pages 52-56. Chung-An Wang, Yen-Chun Liu, Shin-Yi Du, Chio-Wen Lin, Hua-Wen Fu
- **The family 52 β -xylosidase from *Geobacillus stearothermophilus* is a dimer: Structural and biophysical characterization of a glycoside hydrolase.** *Biochimica et Biophysica Acta (BBA) - Proteins & Proteomics*, Volume 1784, Issue 12, December 2008, Pages 1924-1934

Papers with applications of SEDPHAT

- **Solution Structure** of the Complex Formed between **Human Complement C3d** and Full-length **Complement Receptor Type 2**. *Journal of Molecular Biology*, Volume 384, Issue 1, 5 December 2008, Pages 137-150. Keying Li, Azubuike I. Okemefuna, Jayesh Gor, Jonathan P. Hannan, Rengasamy Asokan, V. Michael Holers, Stephen J. Perkins
- **Role of tryptophan-208 residue in cytochrome c oxidation** by ascorbate peroxidase from *Leishmania major*-kinetic studies on Trp208Phe mutant and wild type enzyme. *Biochimica et Biophysica Acta (BBA) - Proteins & Proteomics*, Volume 1784, Issue 5, May 2008, Pages 863-871. Rajesh K. Yadav, Subhankar Dolai, Swati Pal, Subrata Adak
- The **dimeric assembly** of *Photobacterium leiognathi* and *Salmonella typhimurium* SodC1 Cu,Zn superoxide dismutases is affected differently by **active site demetallation** and **pH**: An analytical ultracentrifuge study. *Archives of Biochemistry and Biophysics*, Volume 471, Issue 1, 1 March 2008, Pages 77-84. B. Catacchio, M. D'Orazio, A. Battistoni, E. Chiancone
- **tRNA-dependent asparagine formation** in prokaryotes: Characterization, isolation and structural and functional analysis of a ribonucleoprotein particle generating Asn-tRNA^{Asn}. *Methods*, Volume 44, Issue 2, February 2008, Pages 146-163. Marc Bailly, Mickaël Blaise, Hervé Roy, Marzanna Deniziak, Bernard Lorber, Catherine Birck, Hubert D. Becker, Daniel Kern
- **Characterization and Further Stabilization of Designed Ankyrin Repeat Proteins by Combining Molecular Dynamics Simulations and Experiments**. *Journal of Molecular Biology*, Volume 375, Issue 3, 18 January 2008, Pages 837-854. Gianluca Interlandi, Svava K. Wetzel, Giovanni Settanni, Andreas Plückthun, Amedeo Caflisch
- **On the quaternary structure of a C-type lectin** from *Bothrops jararacussu* venom – BJ-32 (BjcuL). *Toxicon*, Volume 52, Issue 8, 15 December 2008, Pages 944-953. F.P. Silva Jr., G.M.C. Alexandre, C.H.I. Ramos, S.G. De-Simone

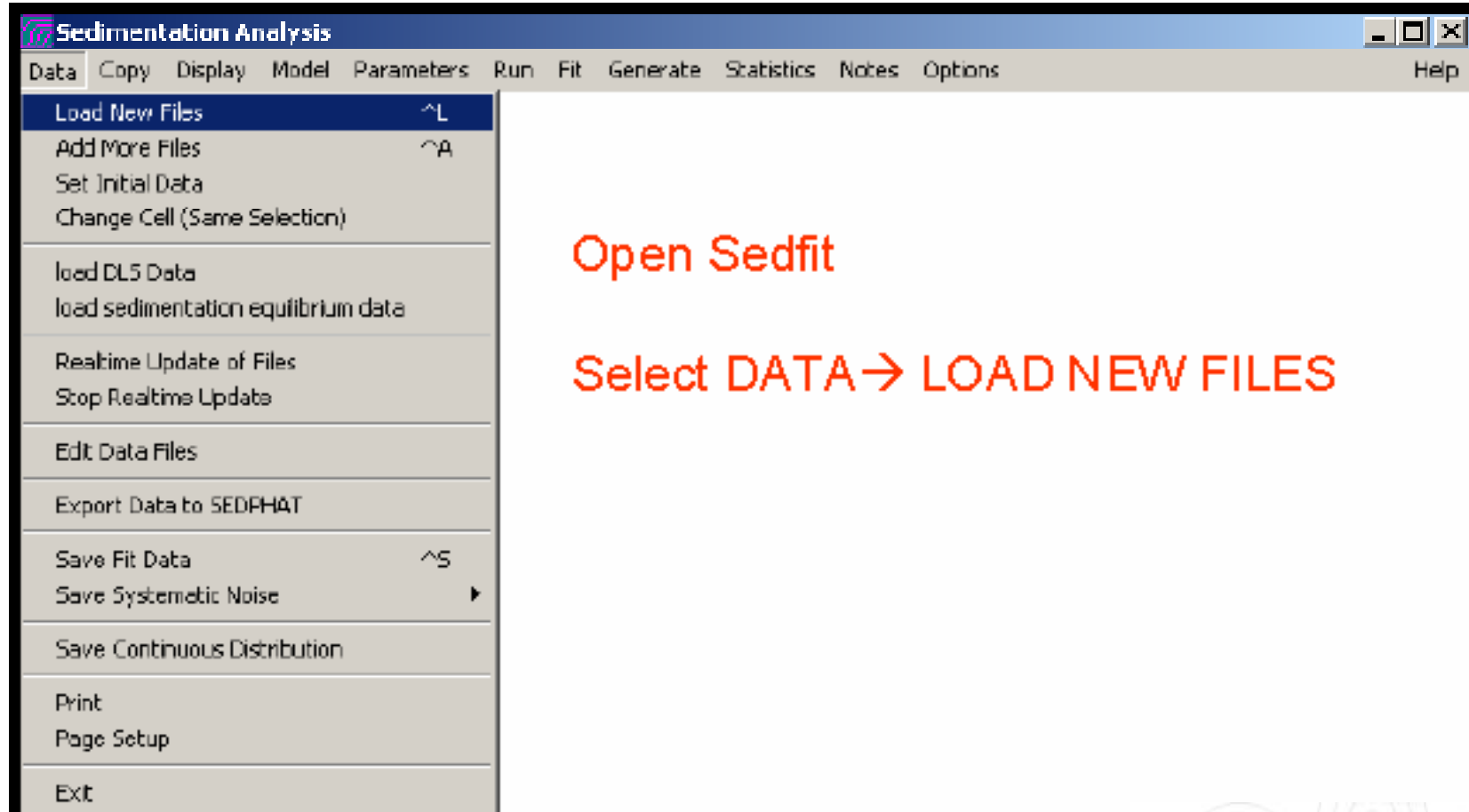
Step by step

Content

- SedFit C(S) analysis
- SedFit C(M) analysis
- SedFit C(f/fo) analysis
- Transform to SEDPHAT
- SEDPHAT models :
 - (1) Single species & heterogenesis
 - (2) Monomer to dimer

Parameter setup

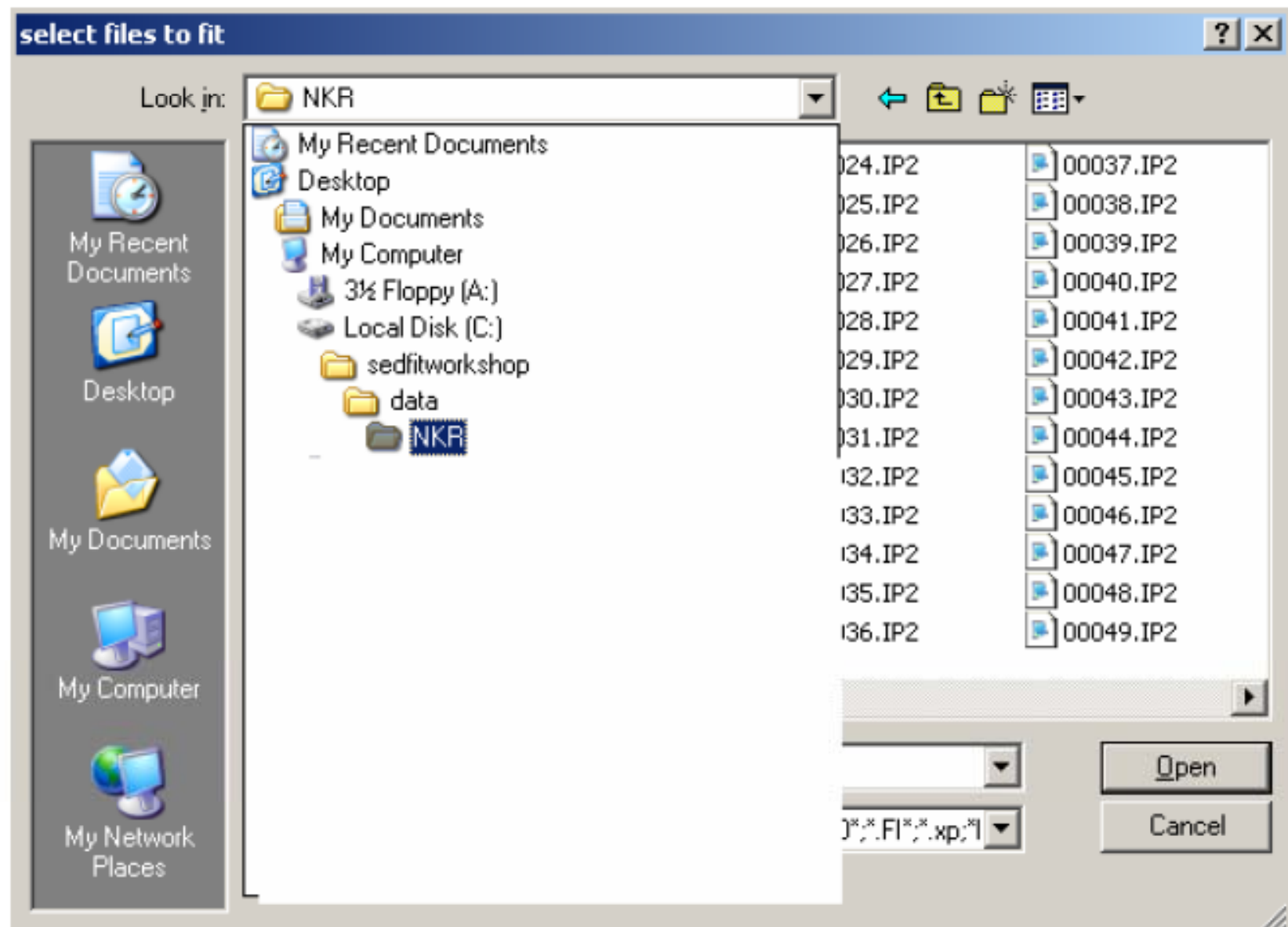
- Buffer density and viscosity
- Protein sample v -bar
- Rmsd
- Z value
- Boundary observation
- Statistics analysis



Open Sedfit

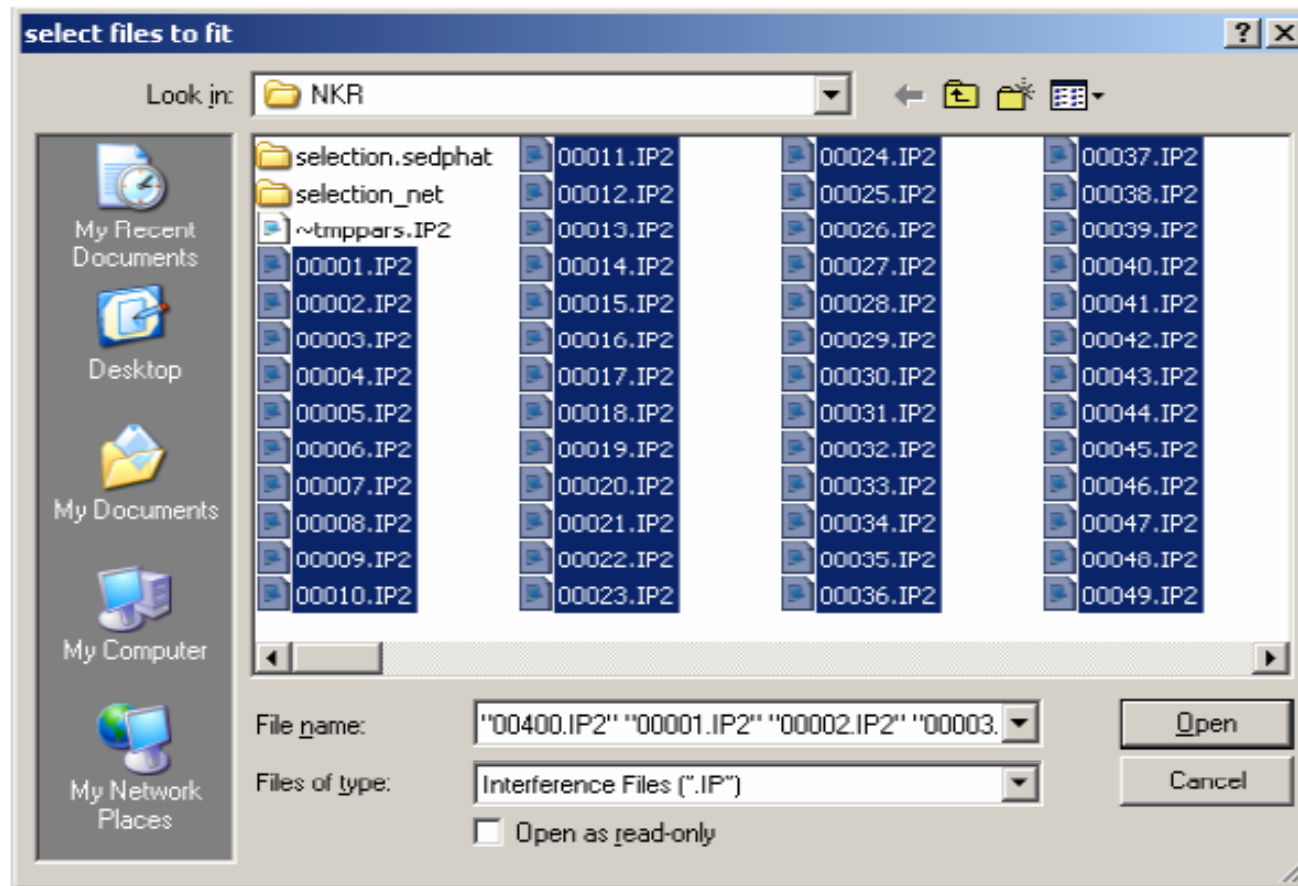
Select DATA → LOAD NEW FILES



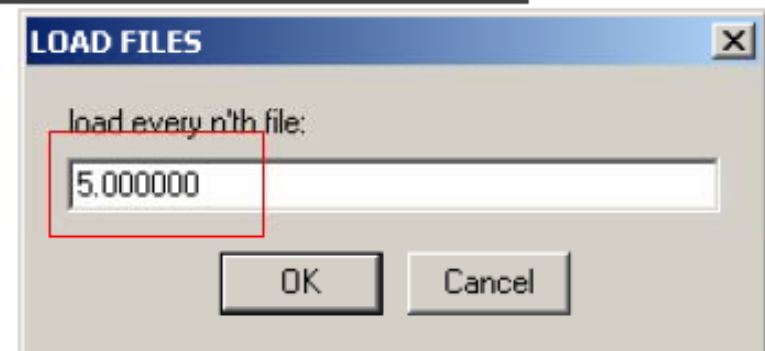


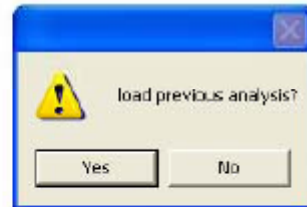
Choose scans within the NKR folder

Select all the interference scans for Cell 2 (highlight scans 1-400)



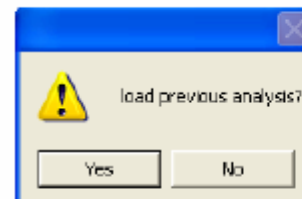
and load every 5th scan





If the data has been loaded before, Sedfit recognizes this and asks if the user wants to use the same parameters as the previous analysis (Sedfit had automatically saved this information in a temporary parameters file “tmppars.ip2” located with the raw data.

Select “No”

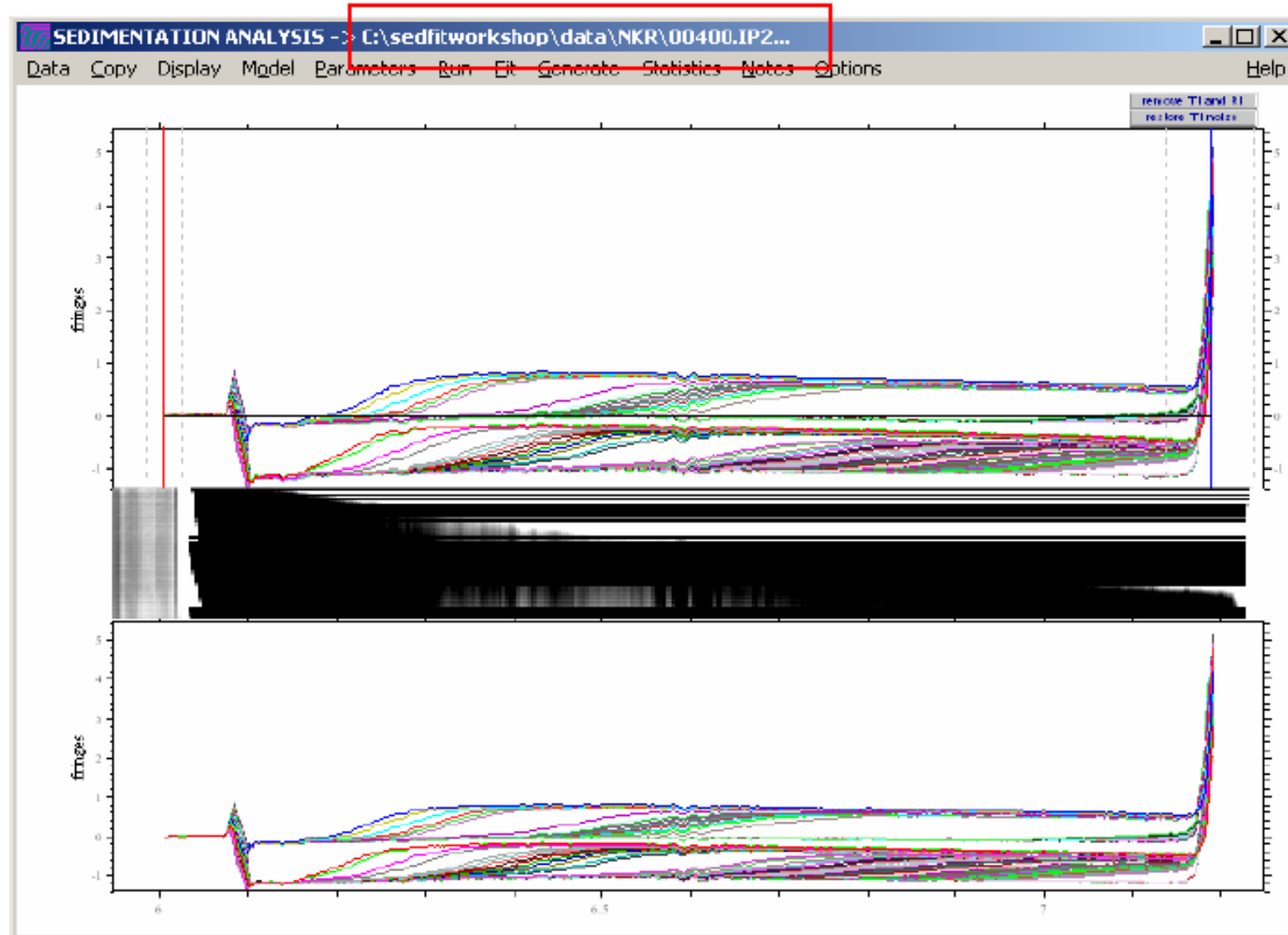


Selecting YES will load all the parameters that we used before (ie. Meniscus, bottom, fitting limits, etc.)

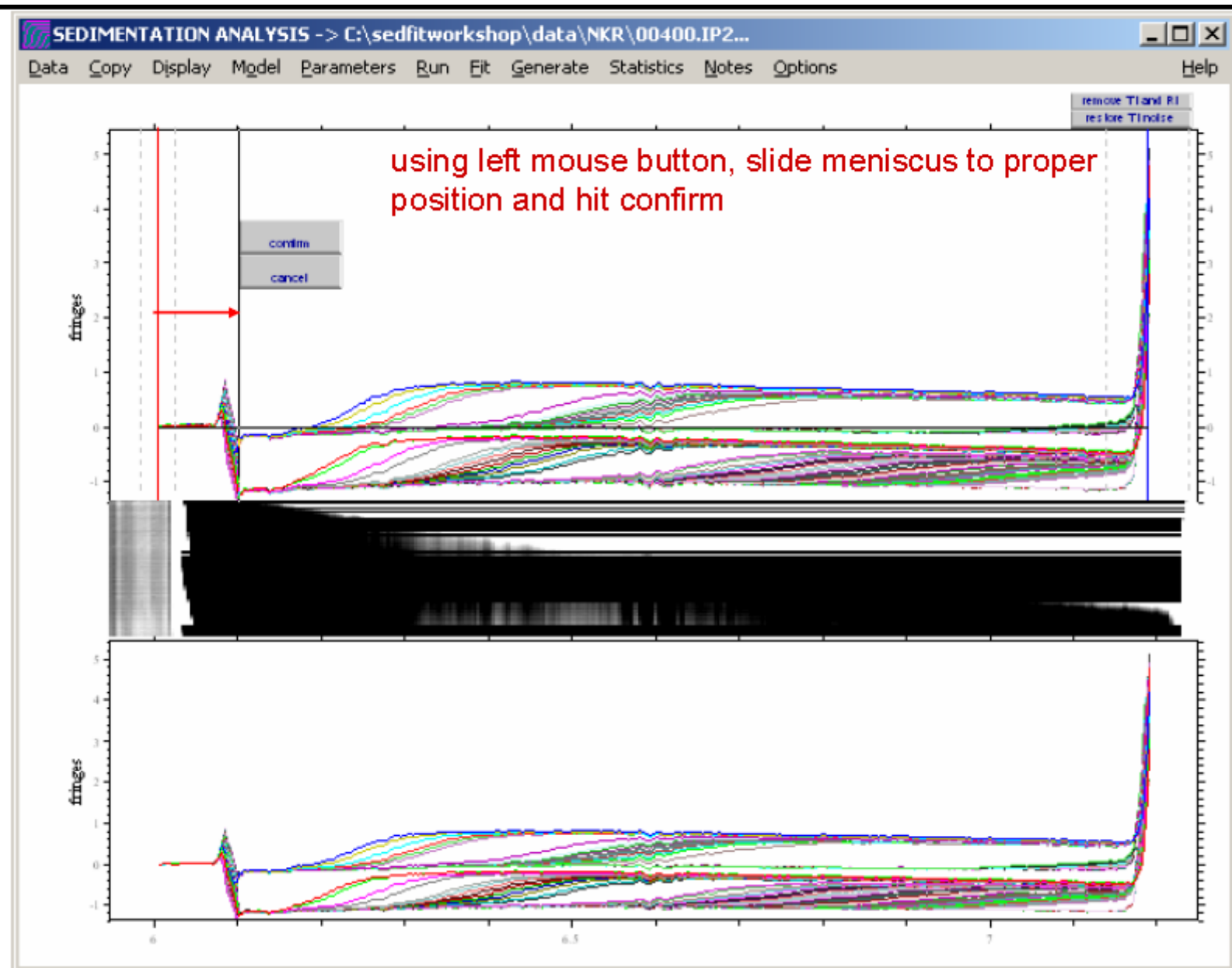
Selecting NO will result in the user needing to set these parameters again.

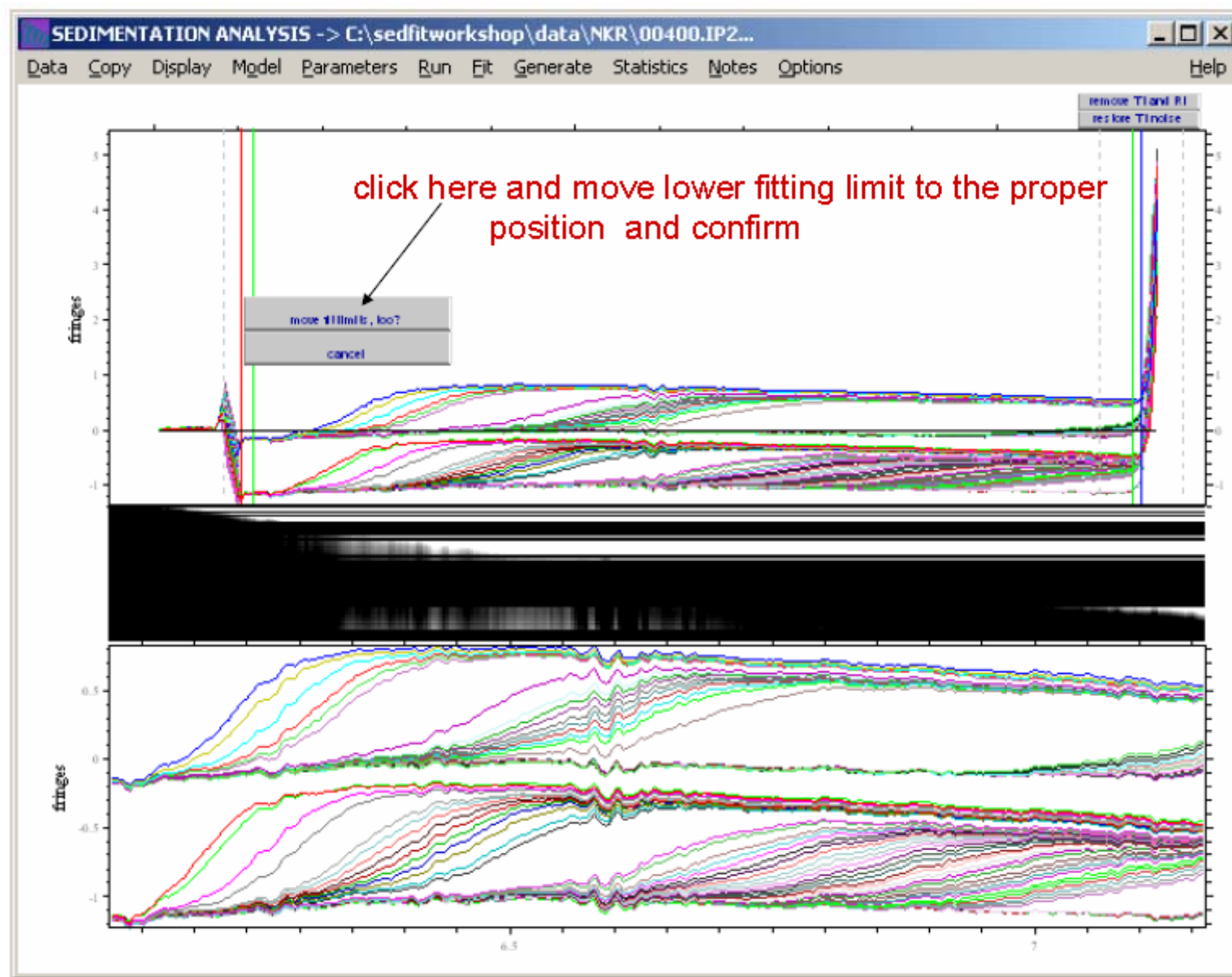
Select YES

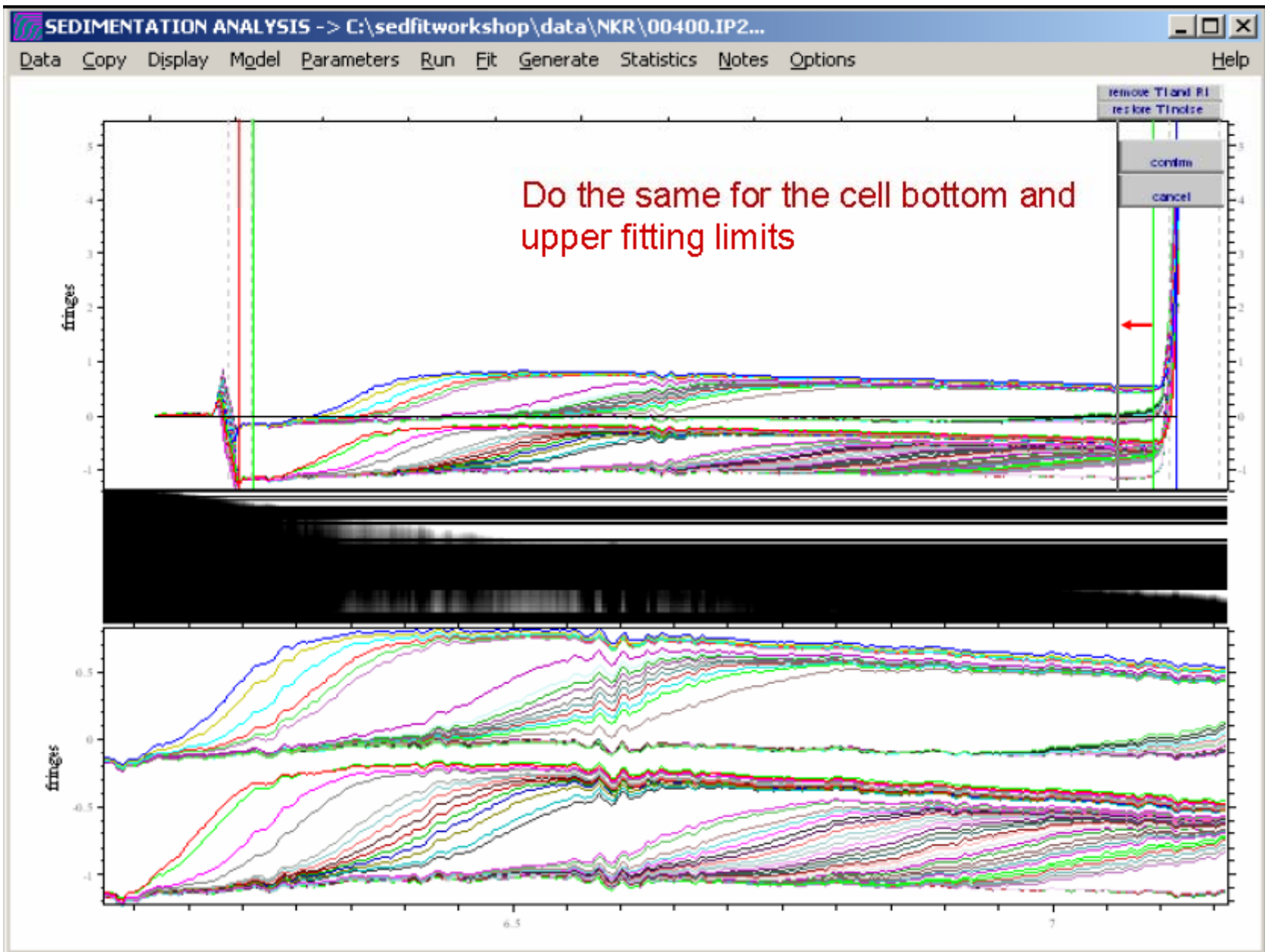
Every 5th scan loaded.



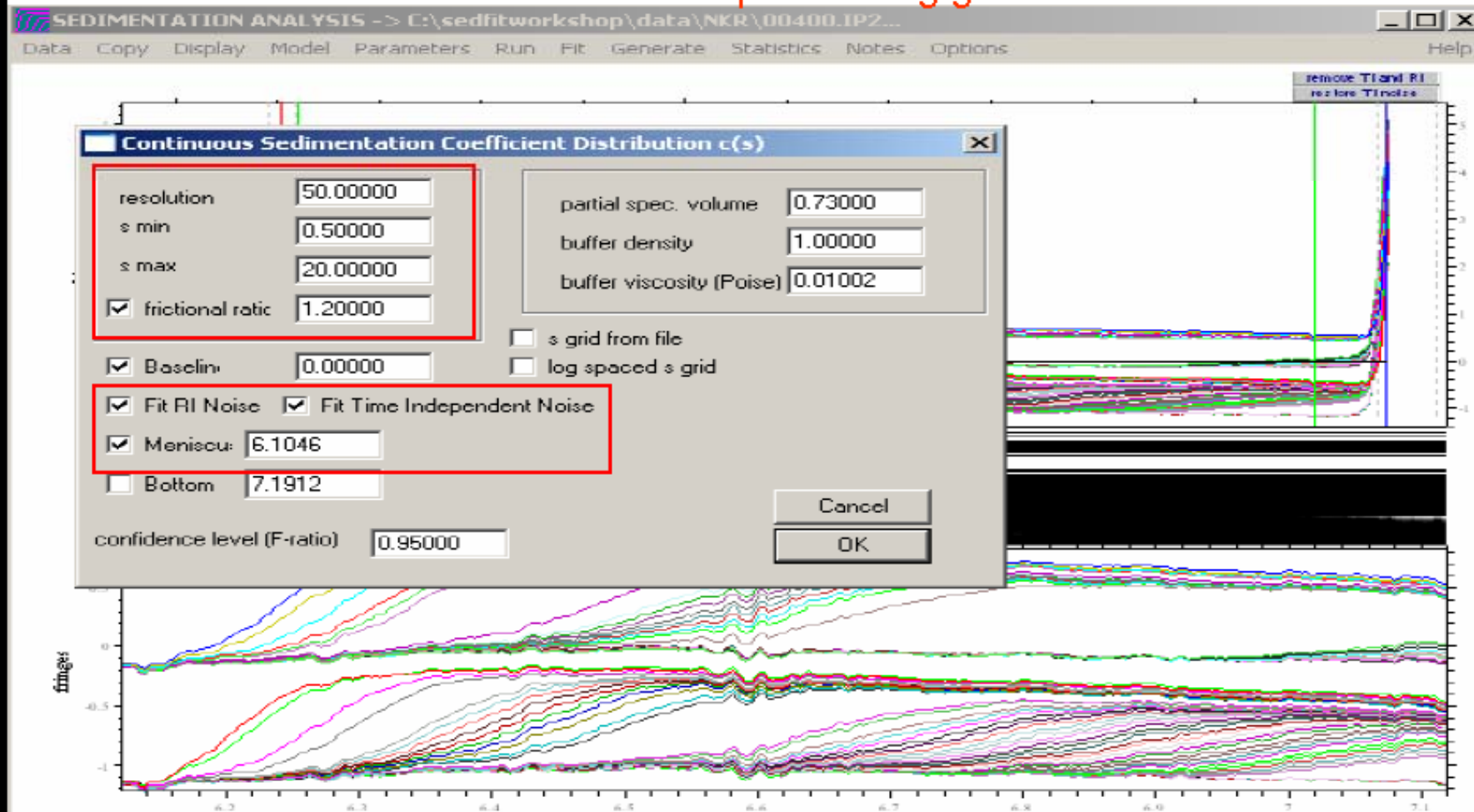
Next, select the meniscus position







Select PARAMETERS and input starting guesses

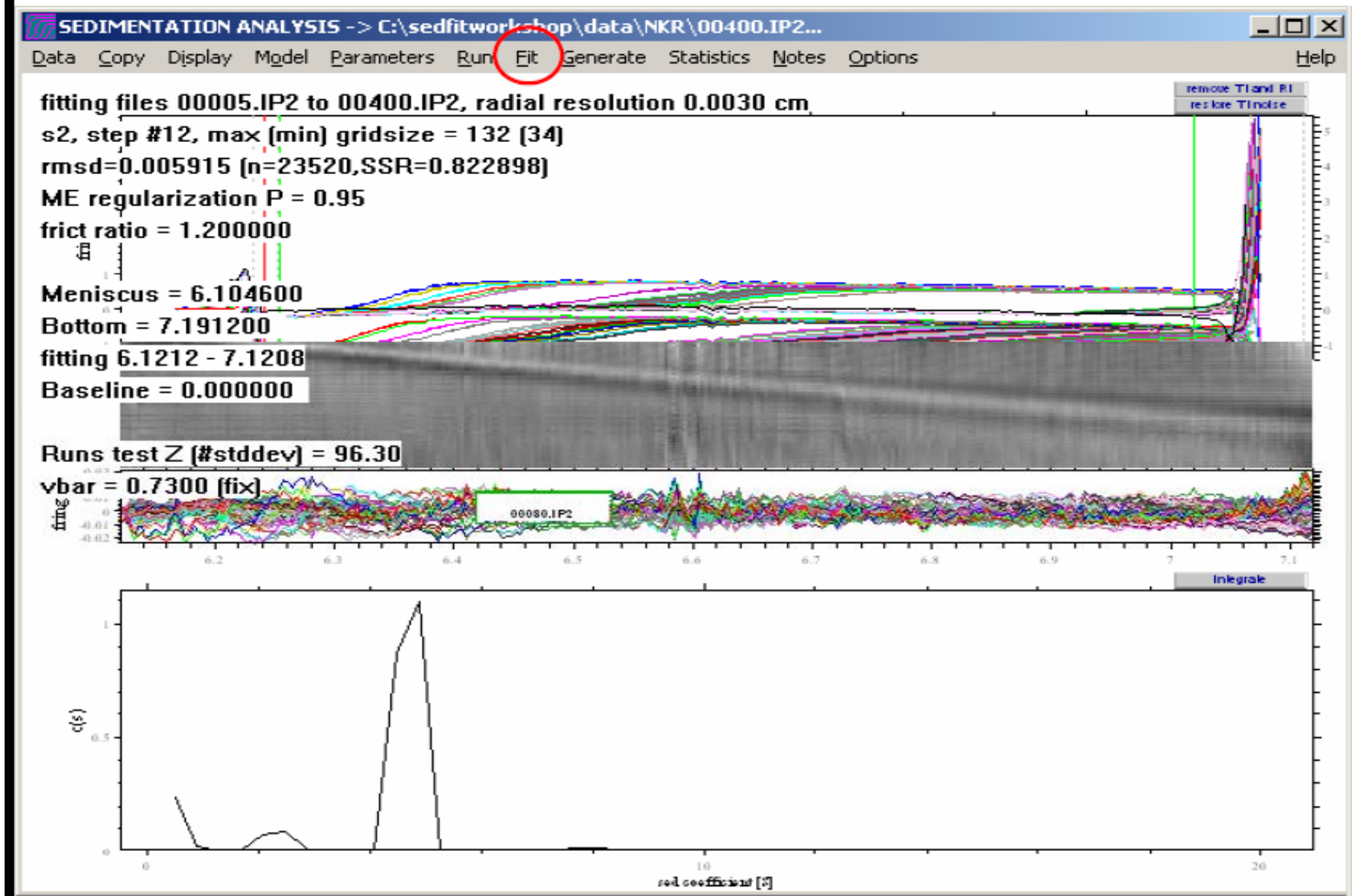


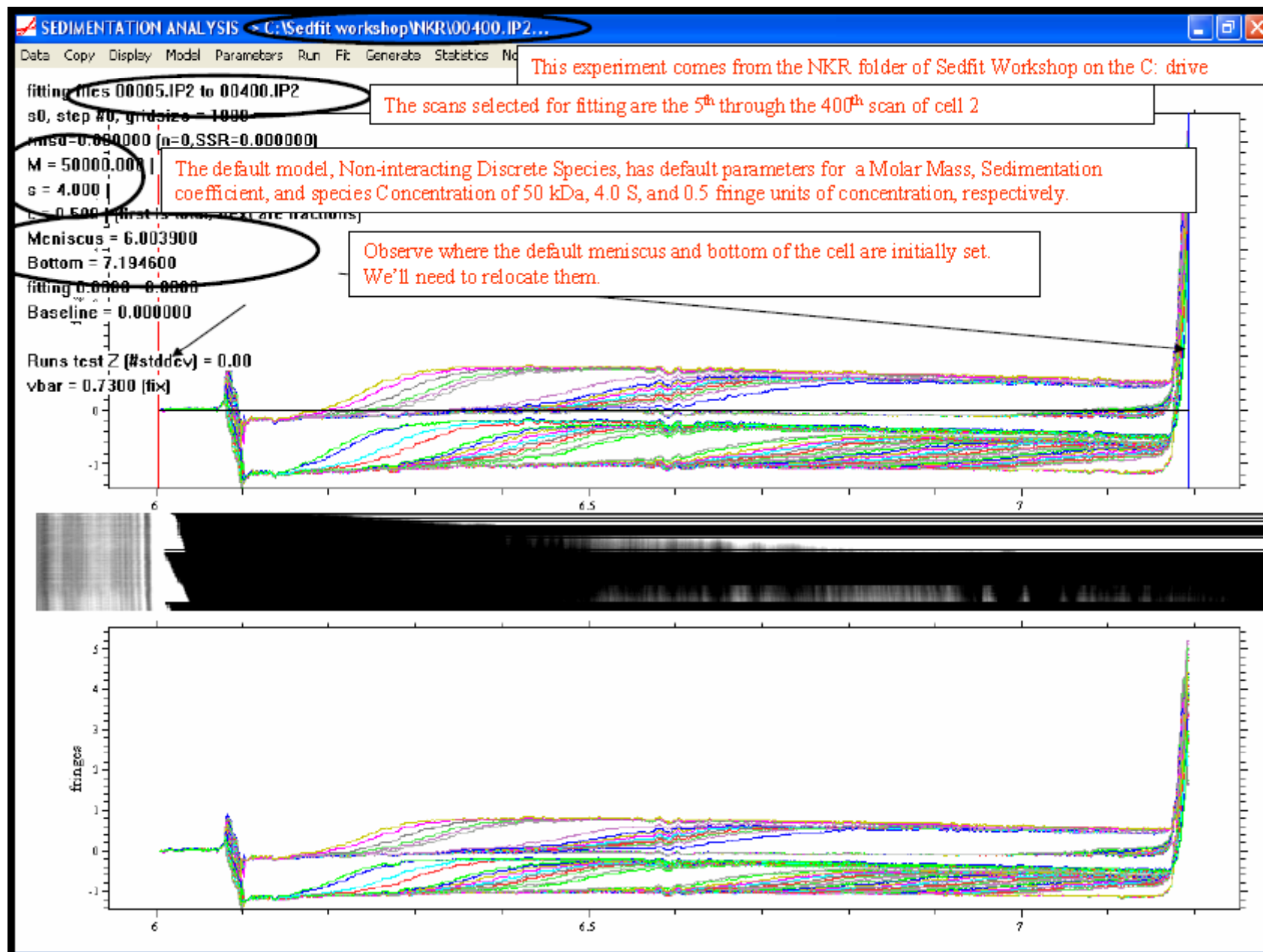
Here we are going to model the experimental data with simulated curves for 50 s-values between 0.5 and 20S.

We will float the frictional ratio Fit for TI and RI noise and Float the meniscus

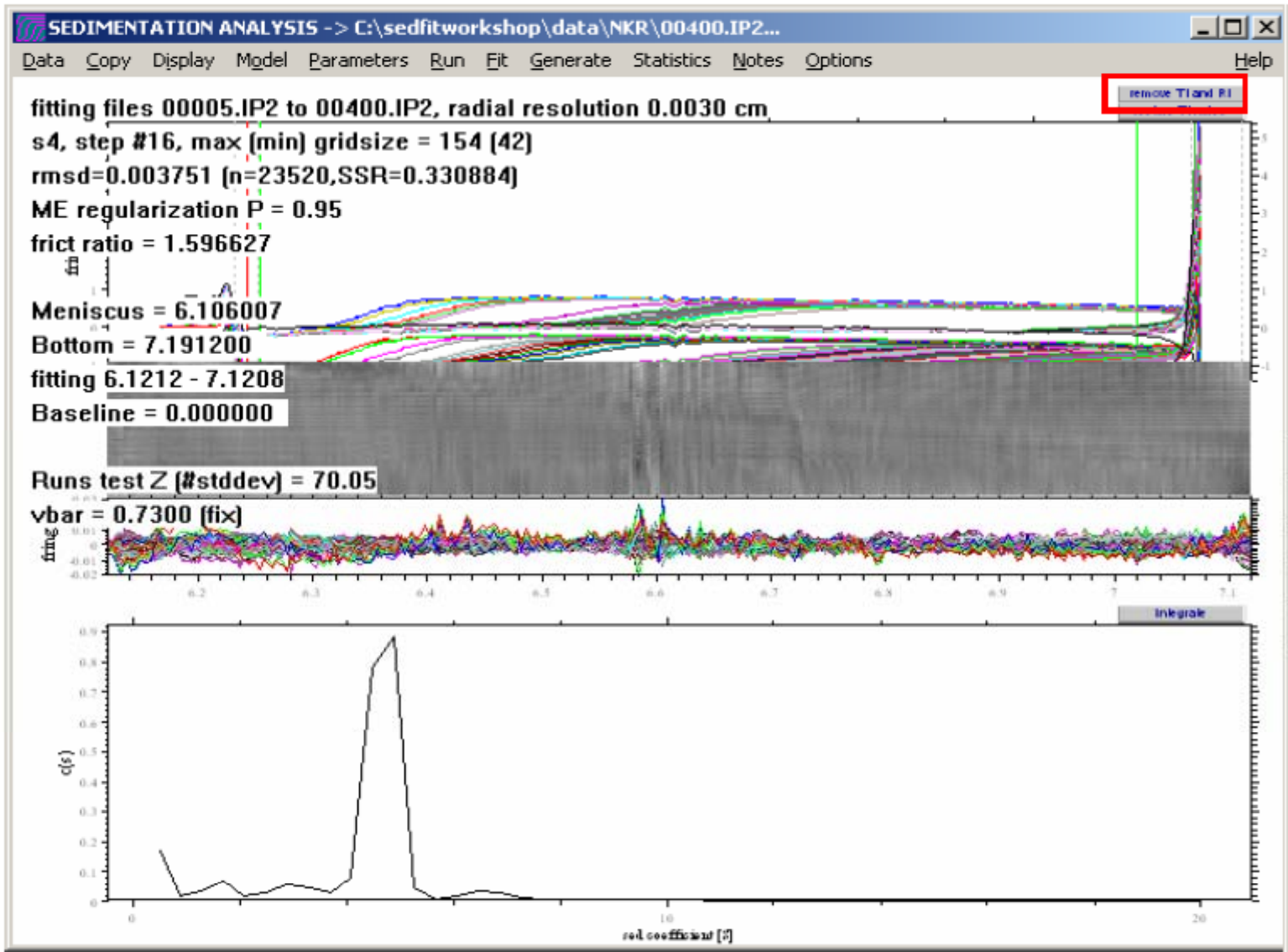
HIT OK

Perform a RUN which optimizes the linear parameters.Then FIT

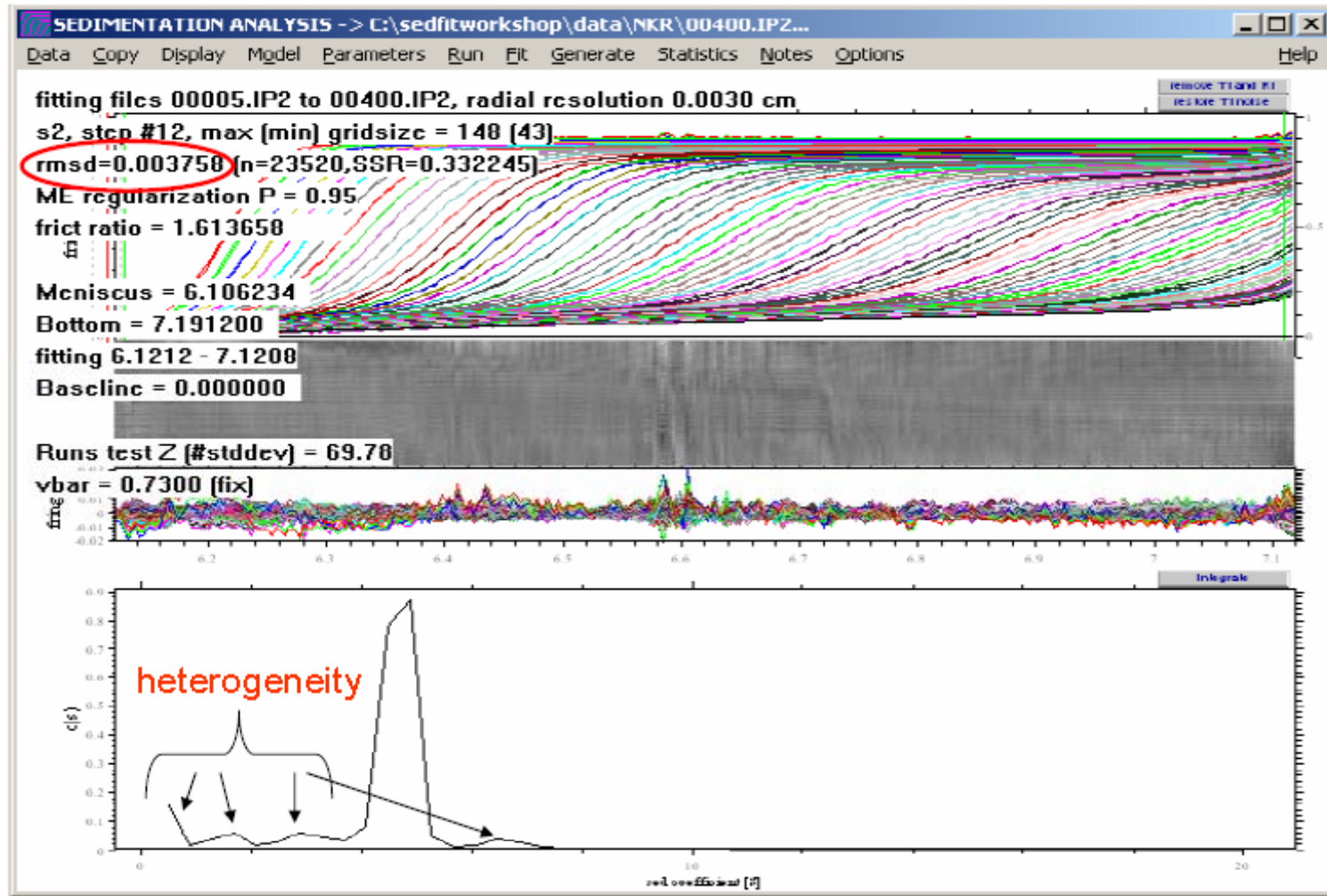


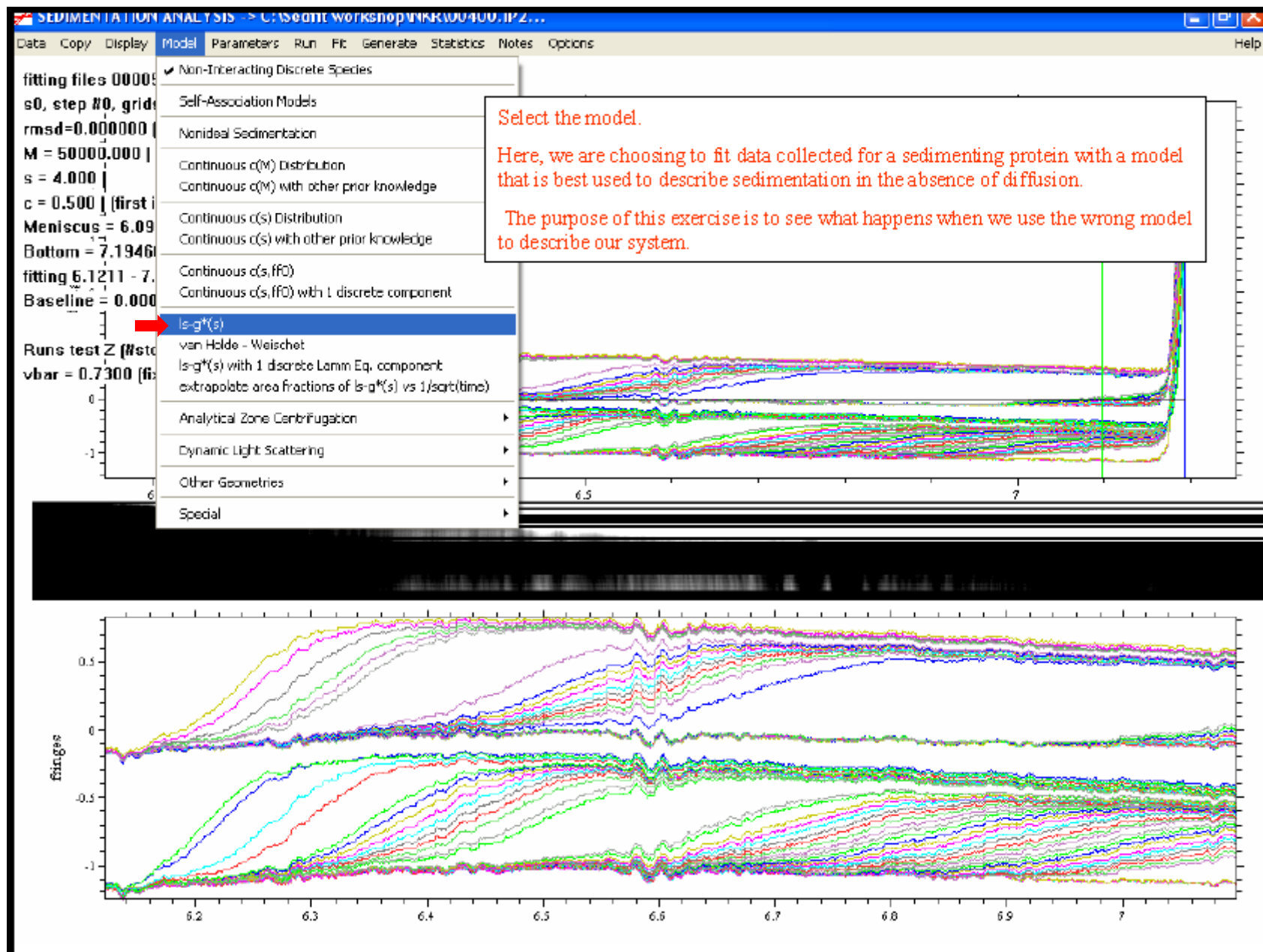


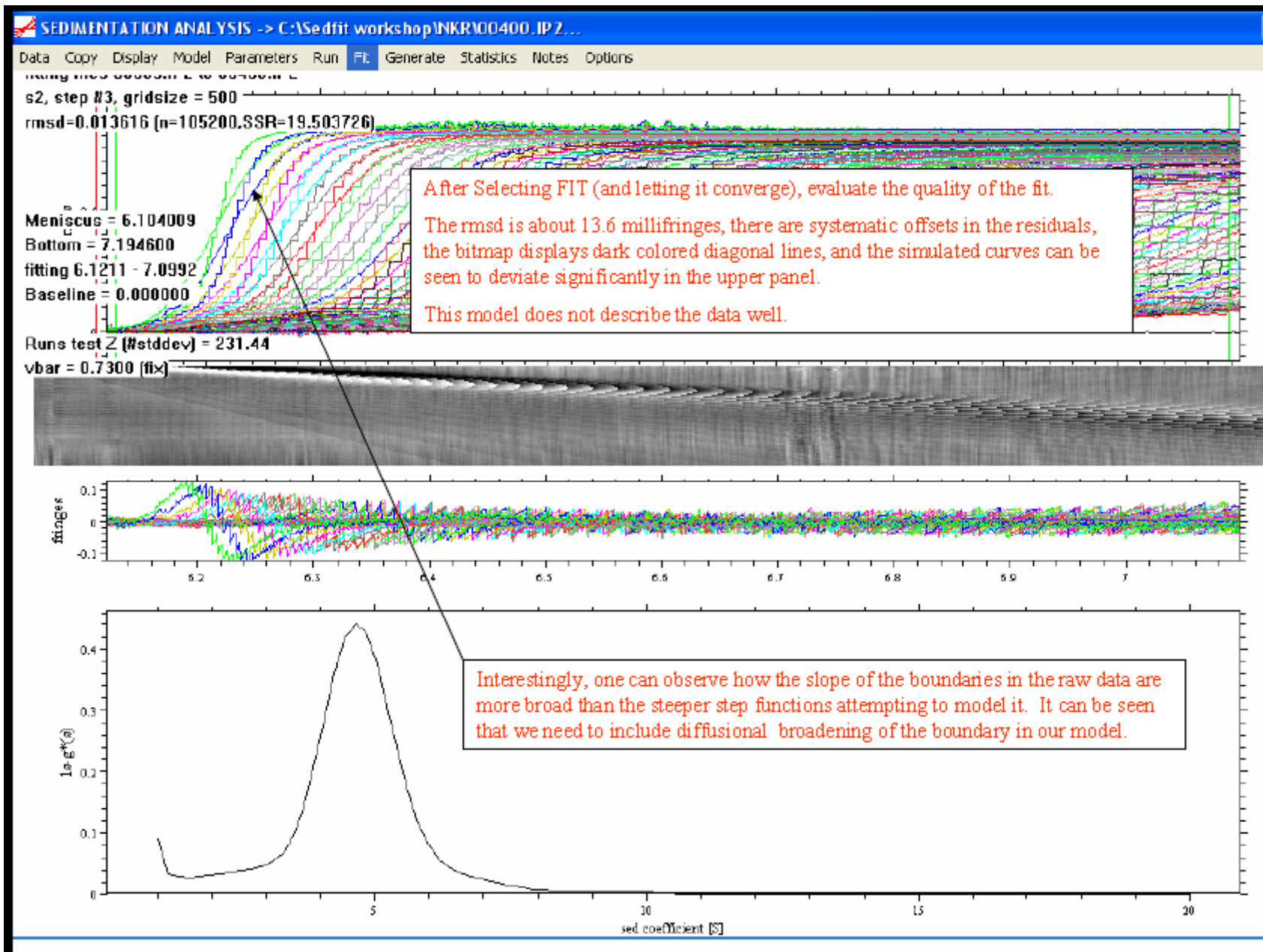
Fitting optimized all parameters. Then subtract the calculated RI and Ti noise.

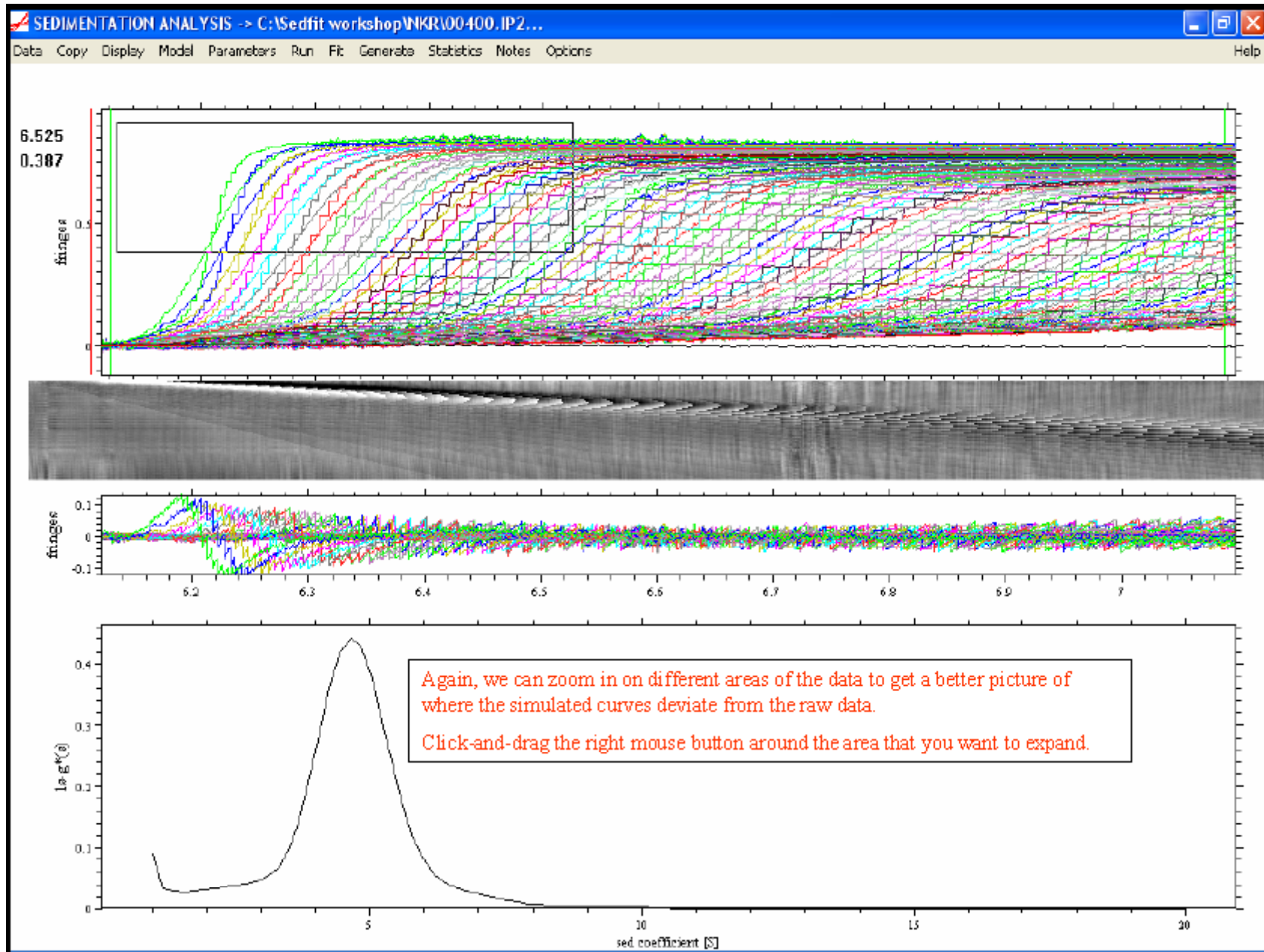


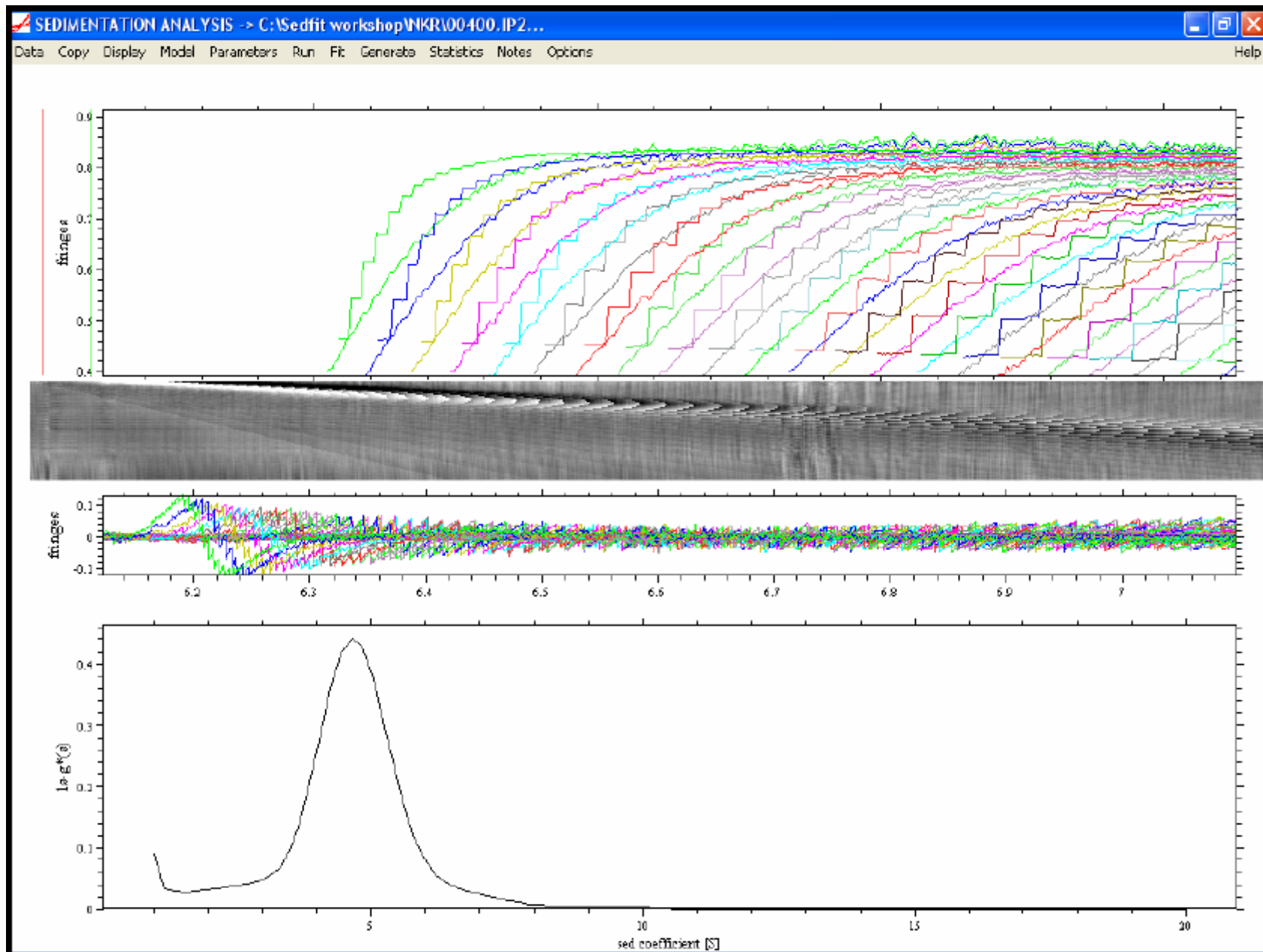
Assess the quality of the fit! Notice root mean square deviation (rmsd) >0.010
Compare to the ls-g*(s) model, and the Non-interacting discrete species model.
Notice the heterogeneity of the sample revealed by the $c(s)$ distribution.

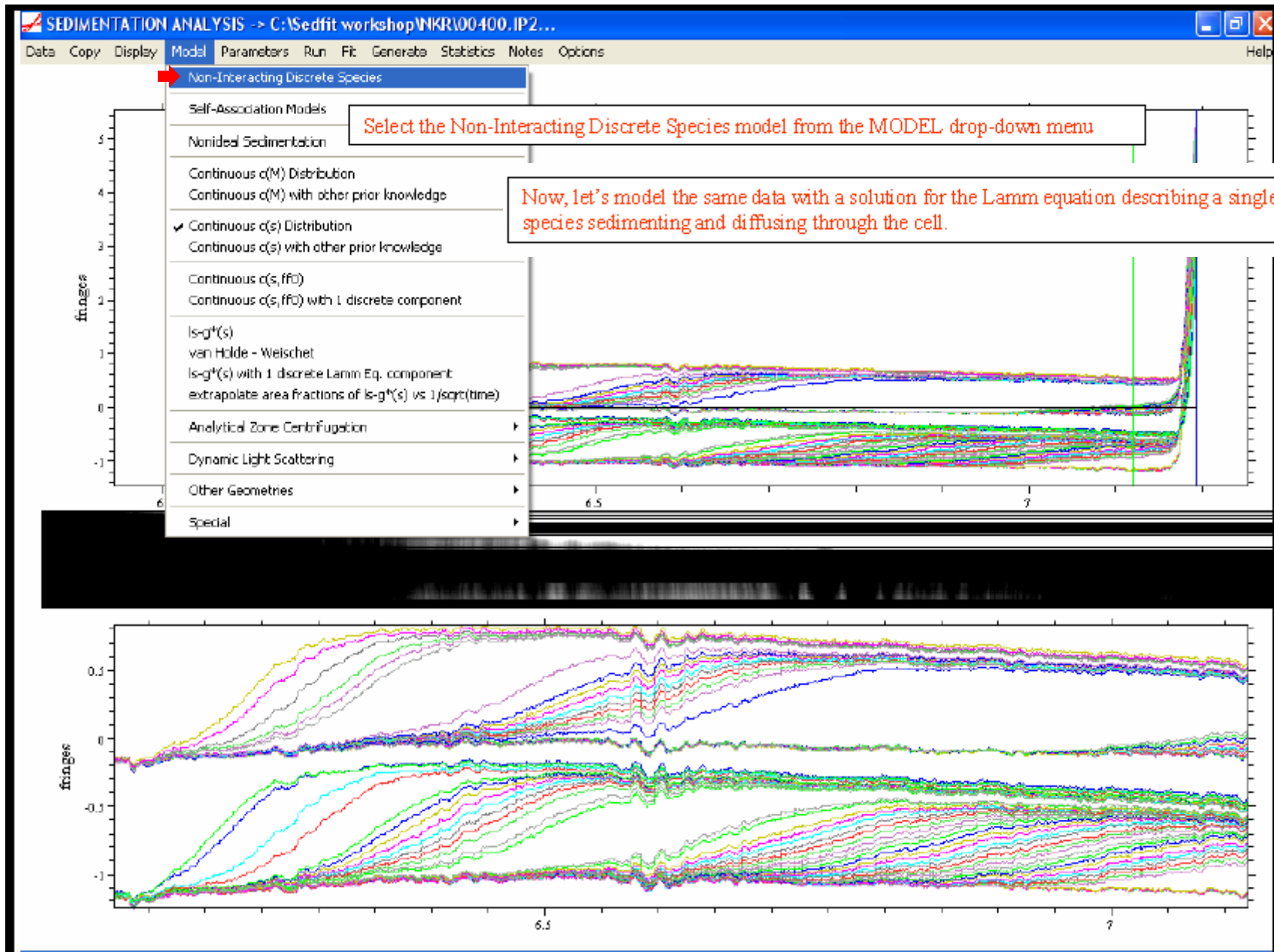


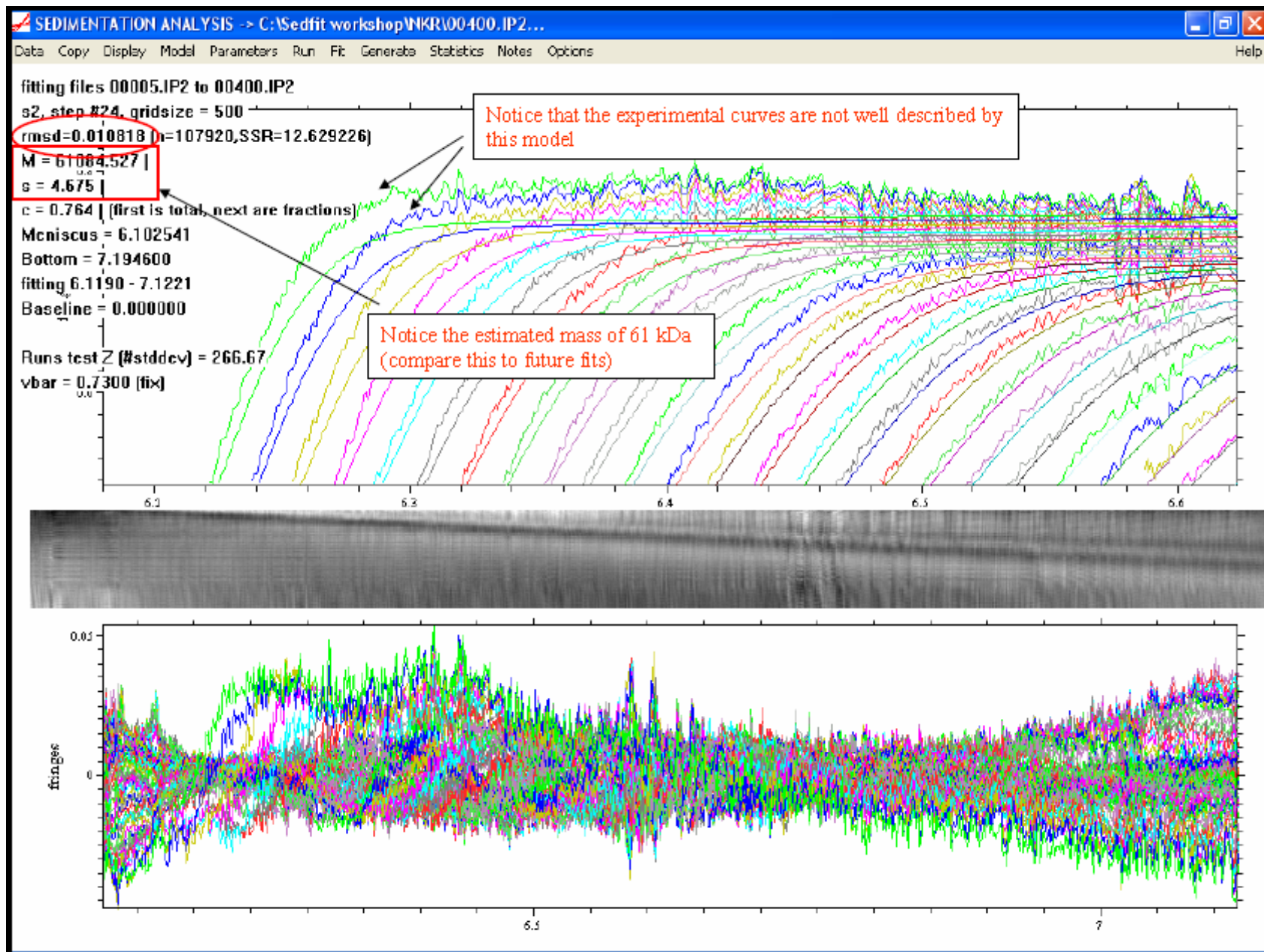




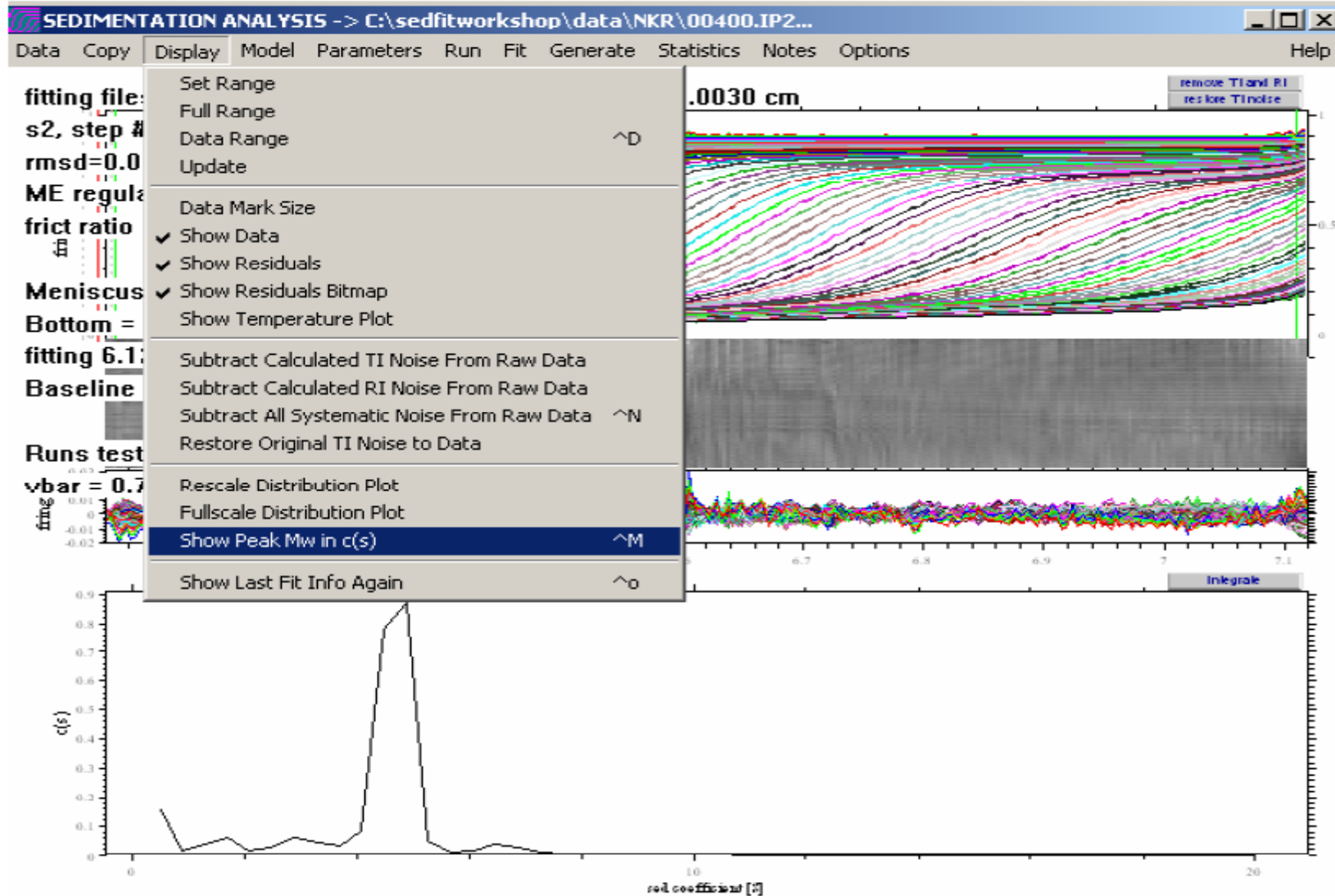




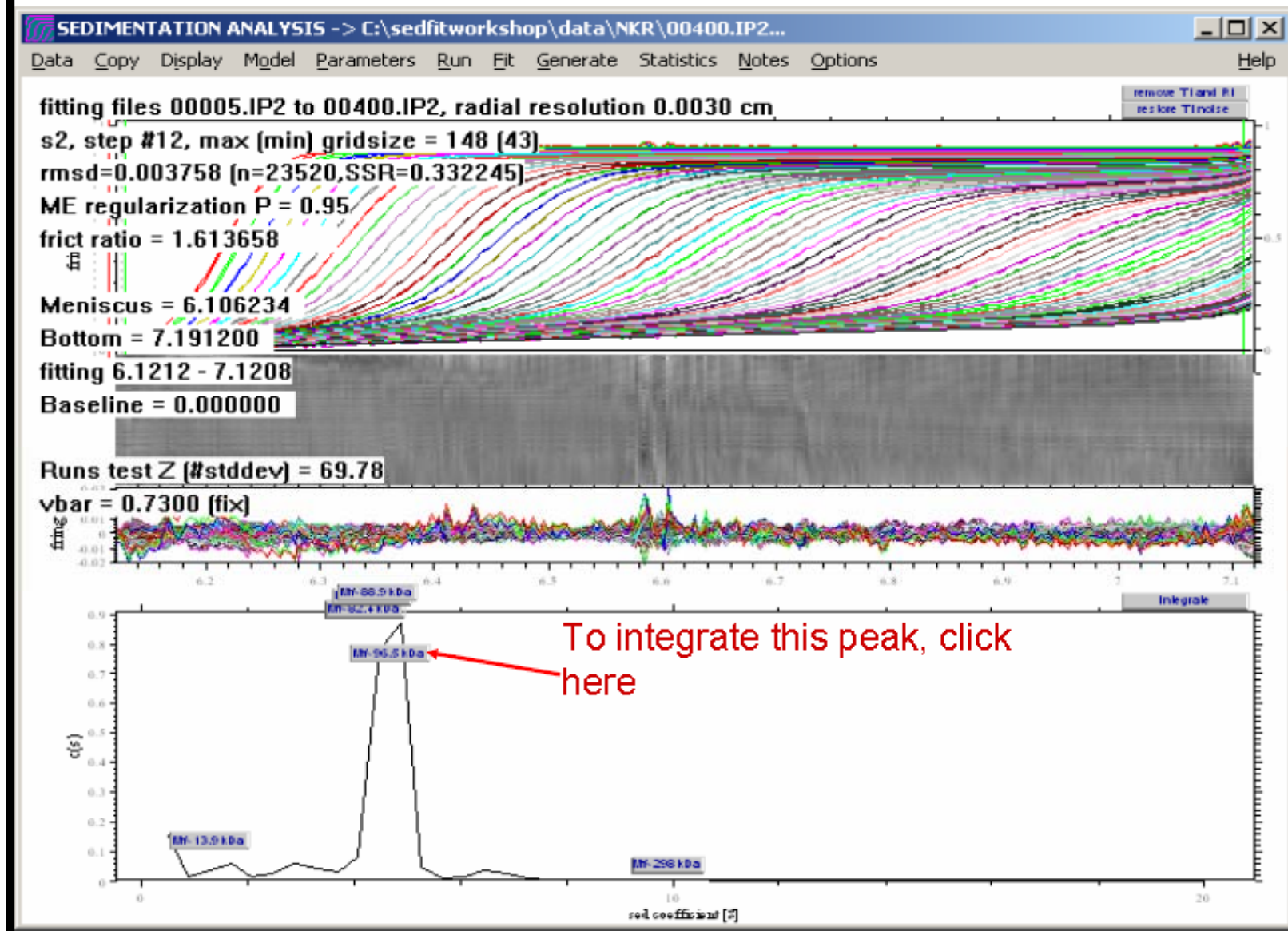




Display the peak MW either from the top menu or by hitting “control M”

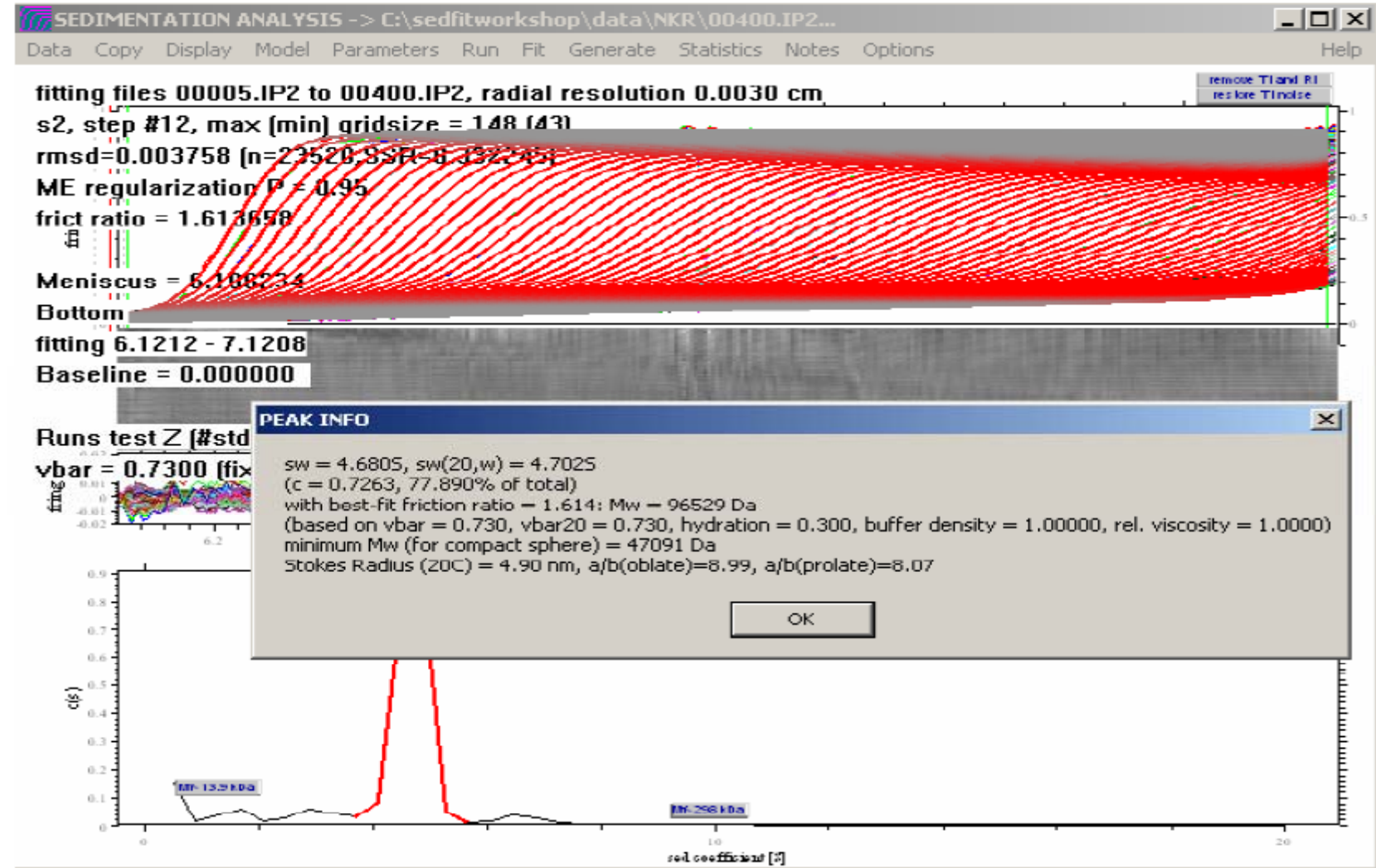


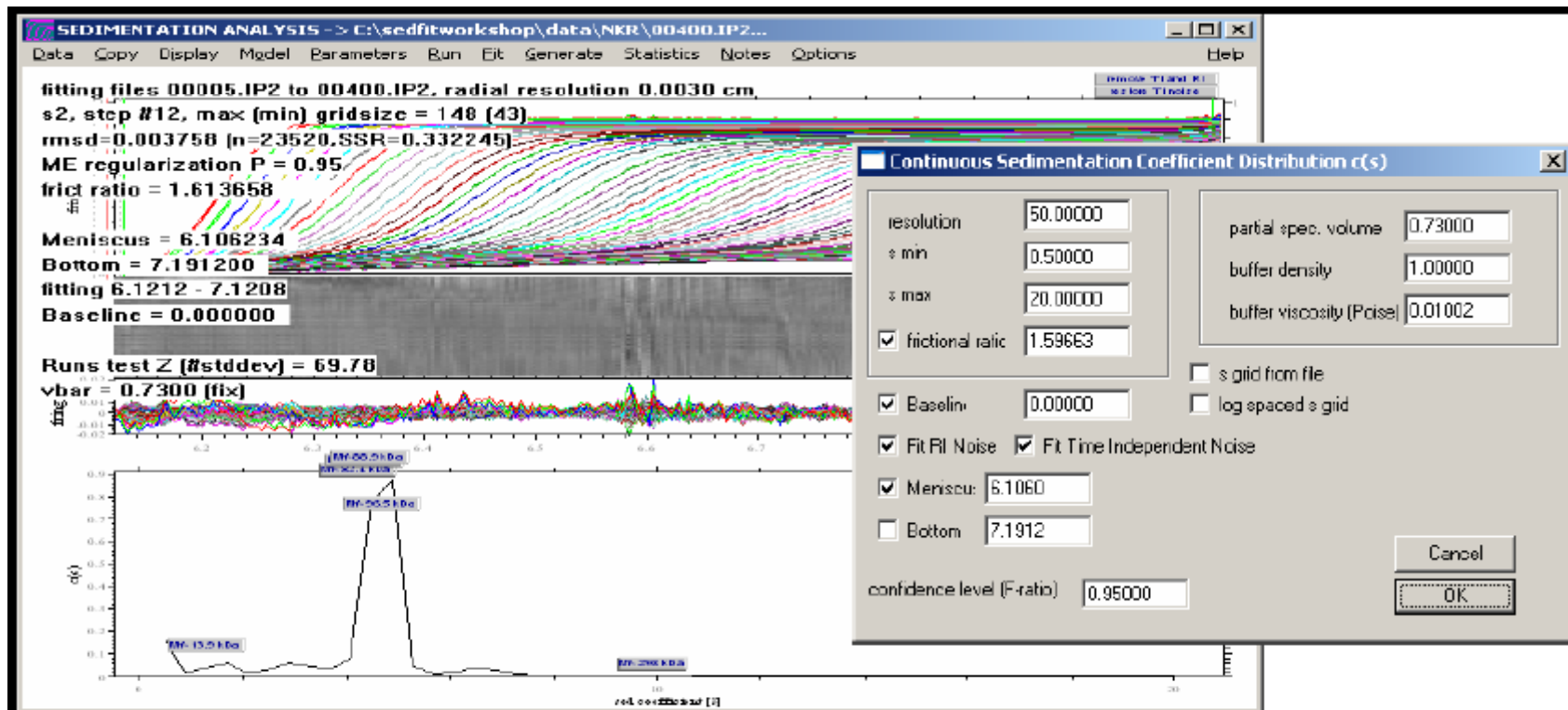
Copy and paste the final cS with MW displayed into Power Point to save



Location of this MW displayed.

Peak information window opens. Copy and paste to ppt.





Try the following:

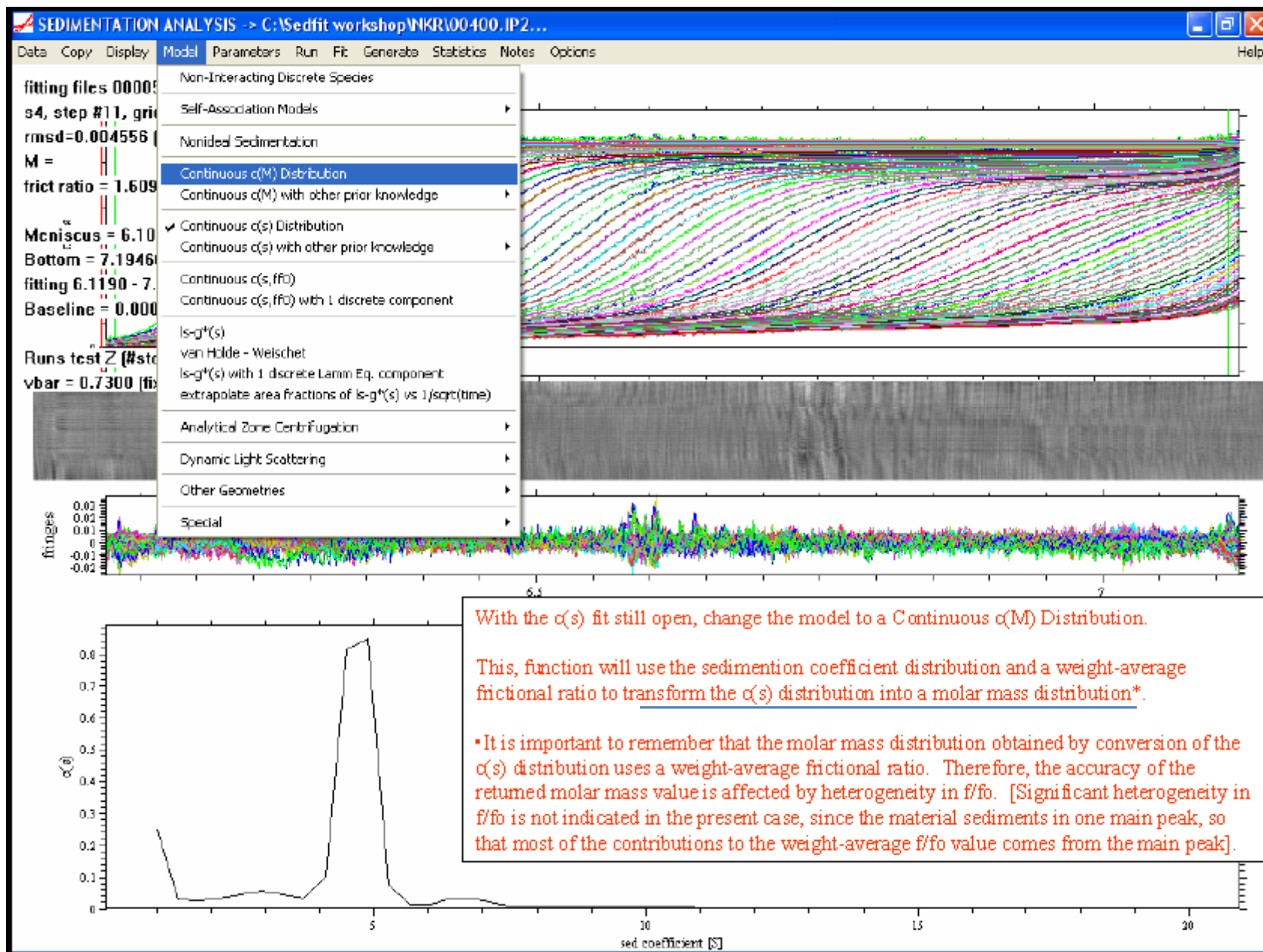
Zoom in the distribution range from 10-20 S (with dragging a rectangle with the right mouse button) – there's not much material out there.

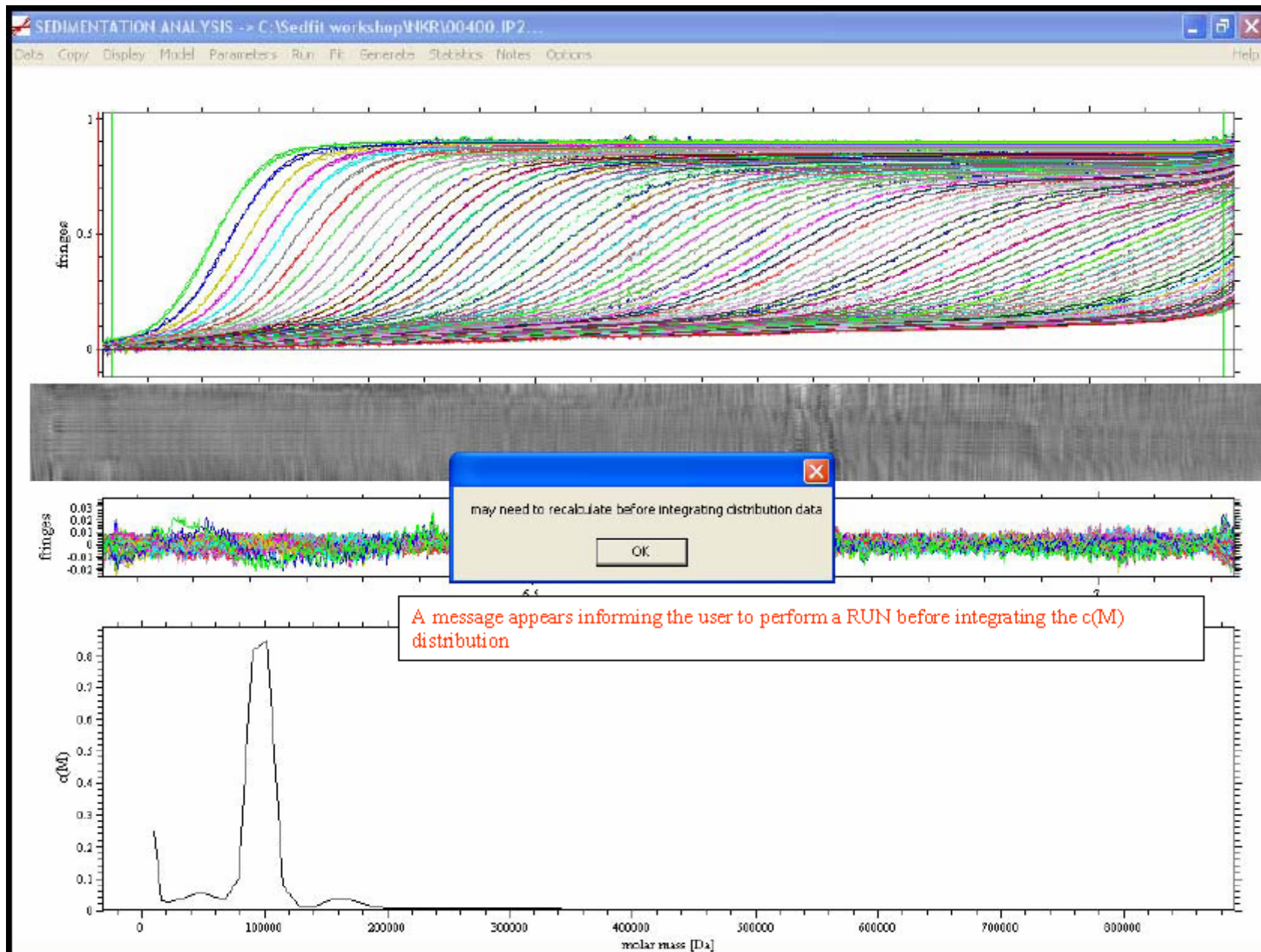
You can further refine the fit by selecting `smax = 15`, and `resolution = 100`

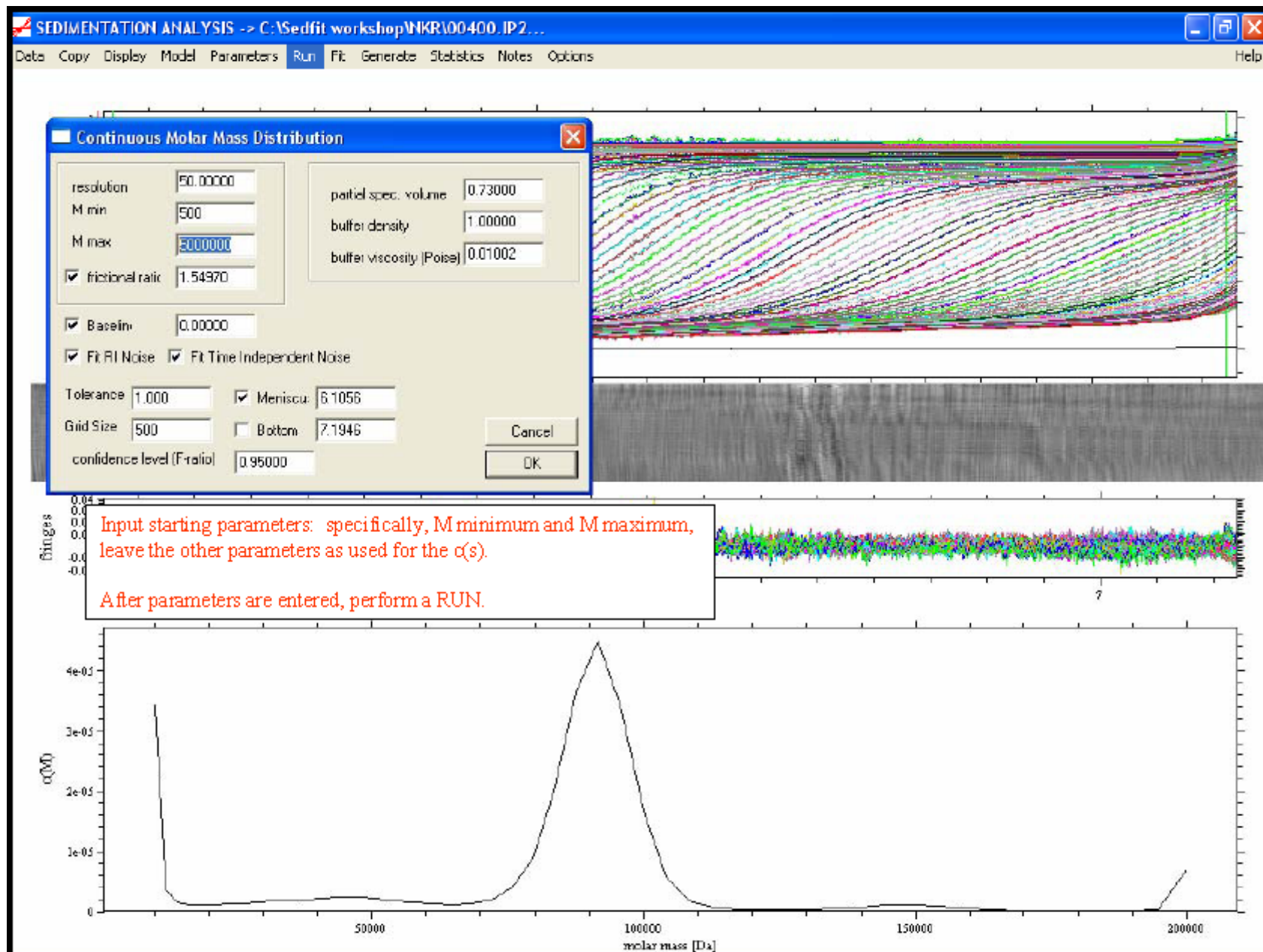
Do a RUN – you'll see a smoother distribution.

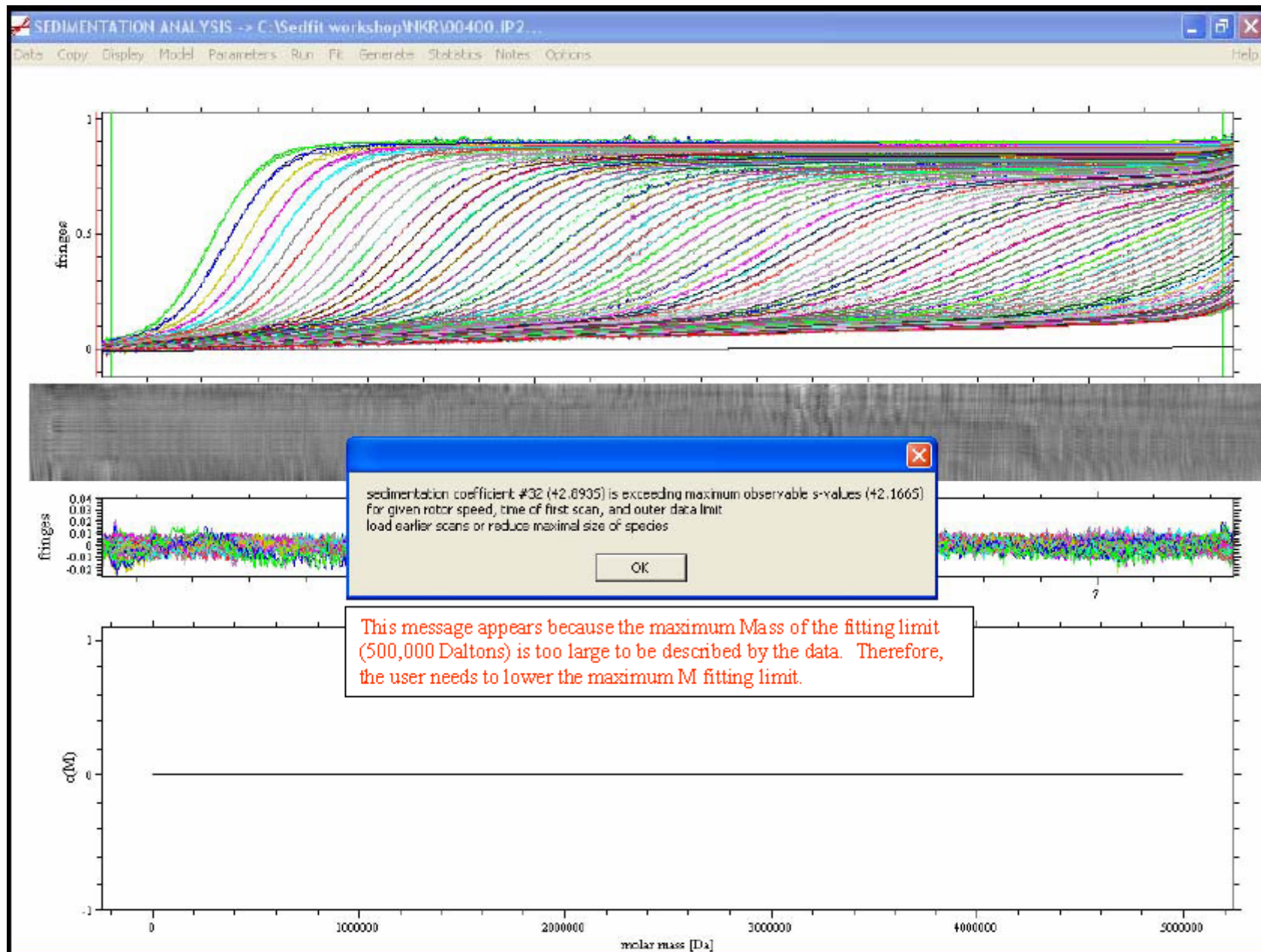
Then,

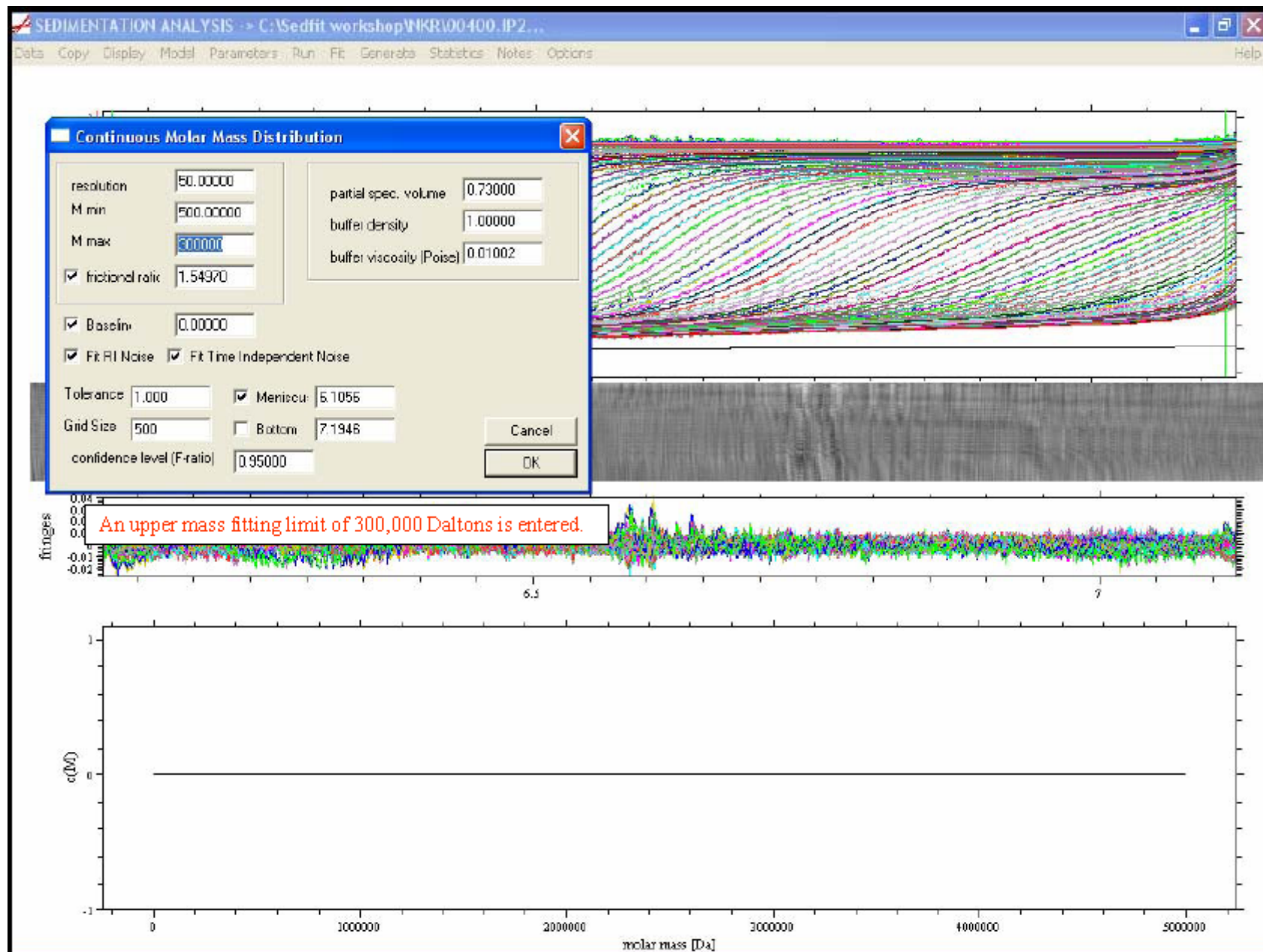
Since the distribution goes up at s-min, test what decreasing the lower limit s-min does (for example, set s-min = 0.2). One of the two should be expected: a) a new peak appears and c(s) goes down at the new s-min, and rmsd gets slightly better; b) c(s) goes up further at s-min. What that means is that s-min is low enough to describe all possible sedimentation at the lower end, and is correlated with the baseline (baseline = no sedimentation). This is no problem. In this case, you can simply ignore this partial peak at s-min, and you can zoom into the rest of the distribution to see that better. The option b) is what happens here.

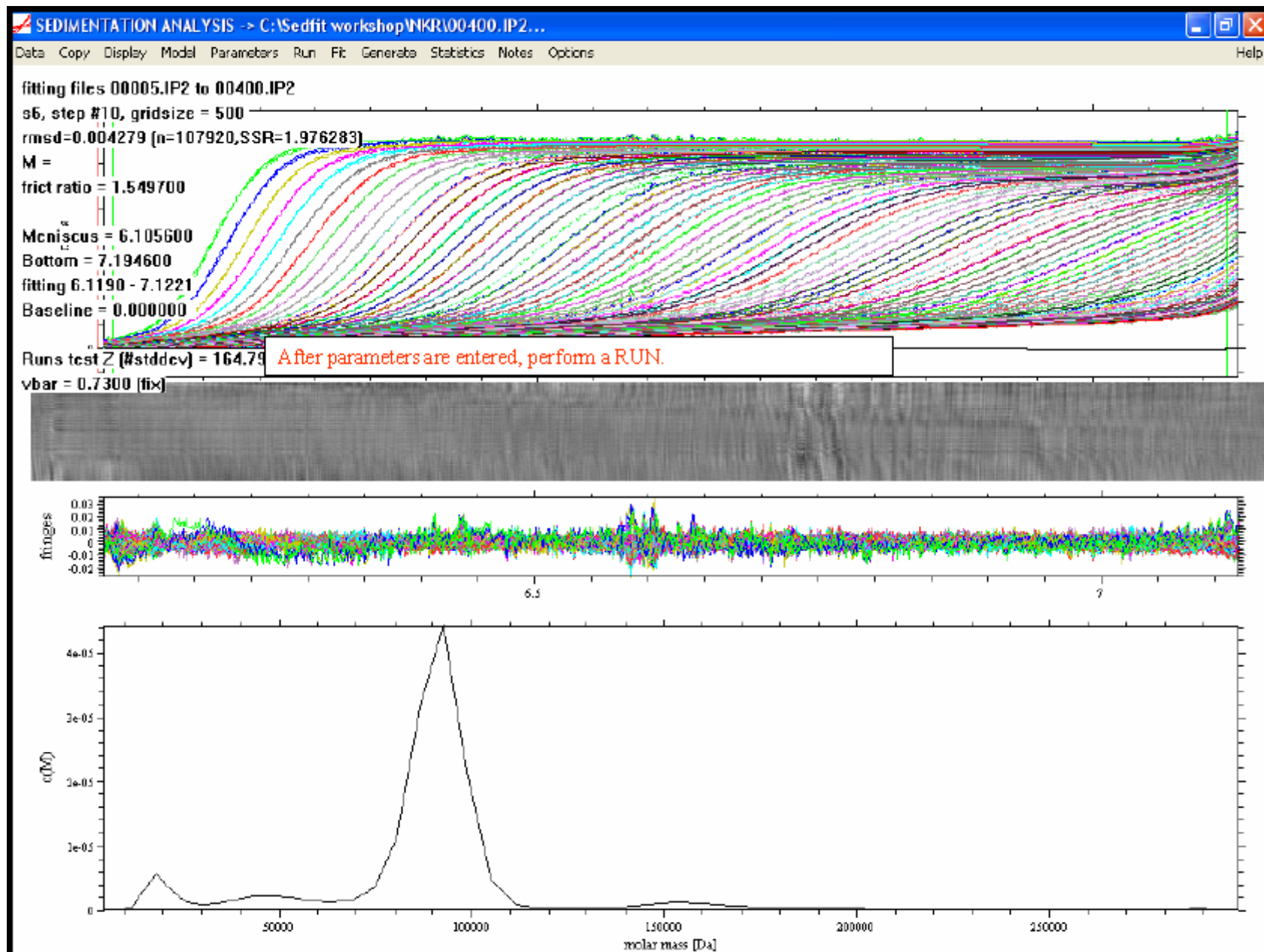


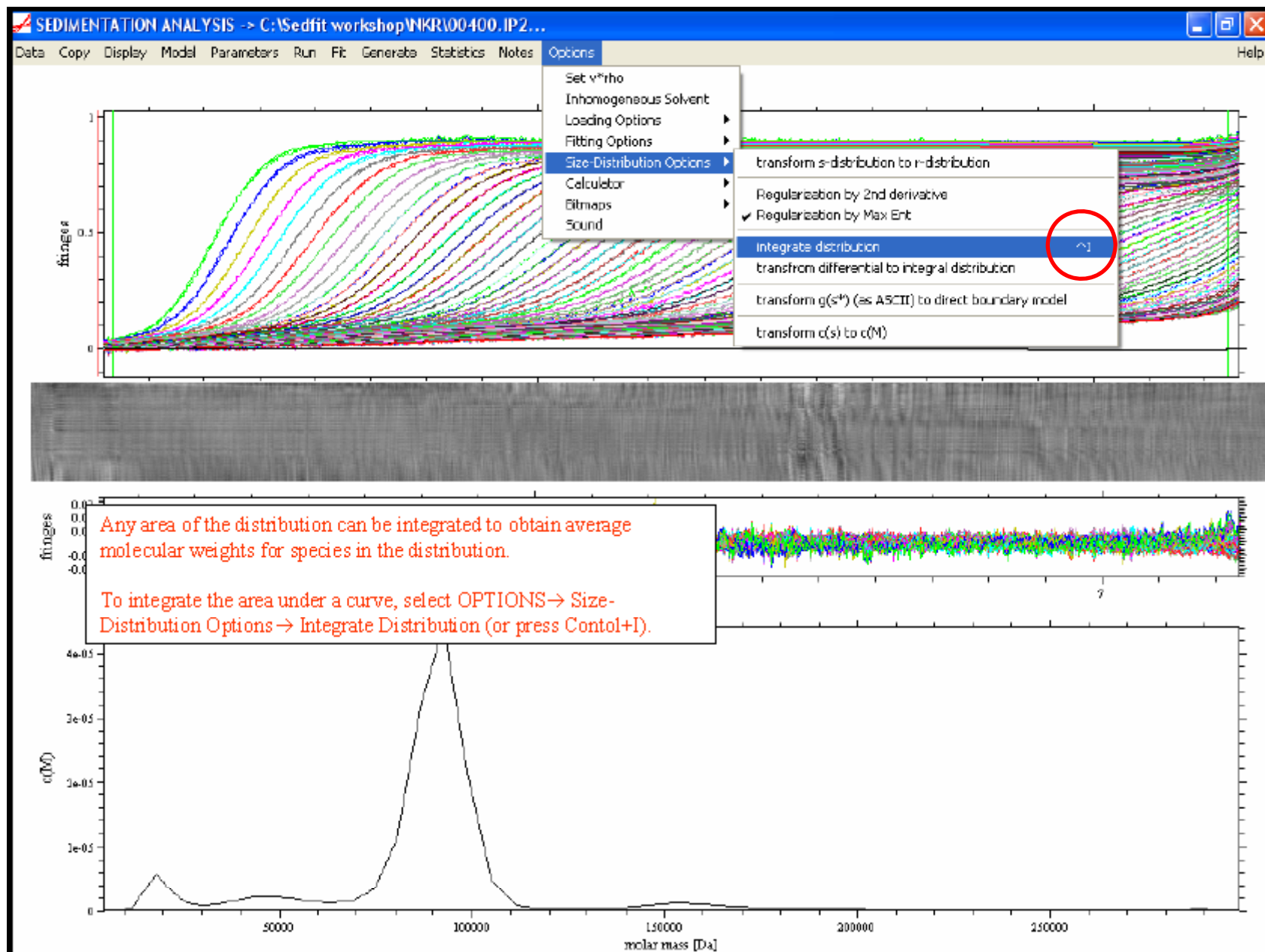


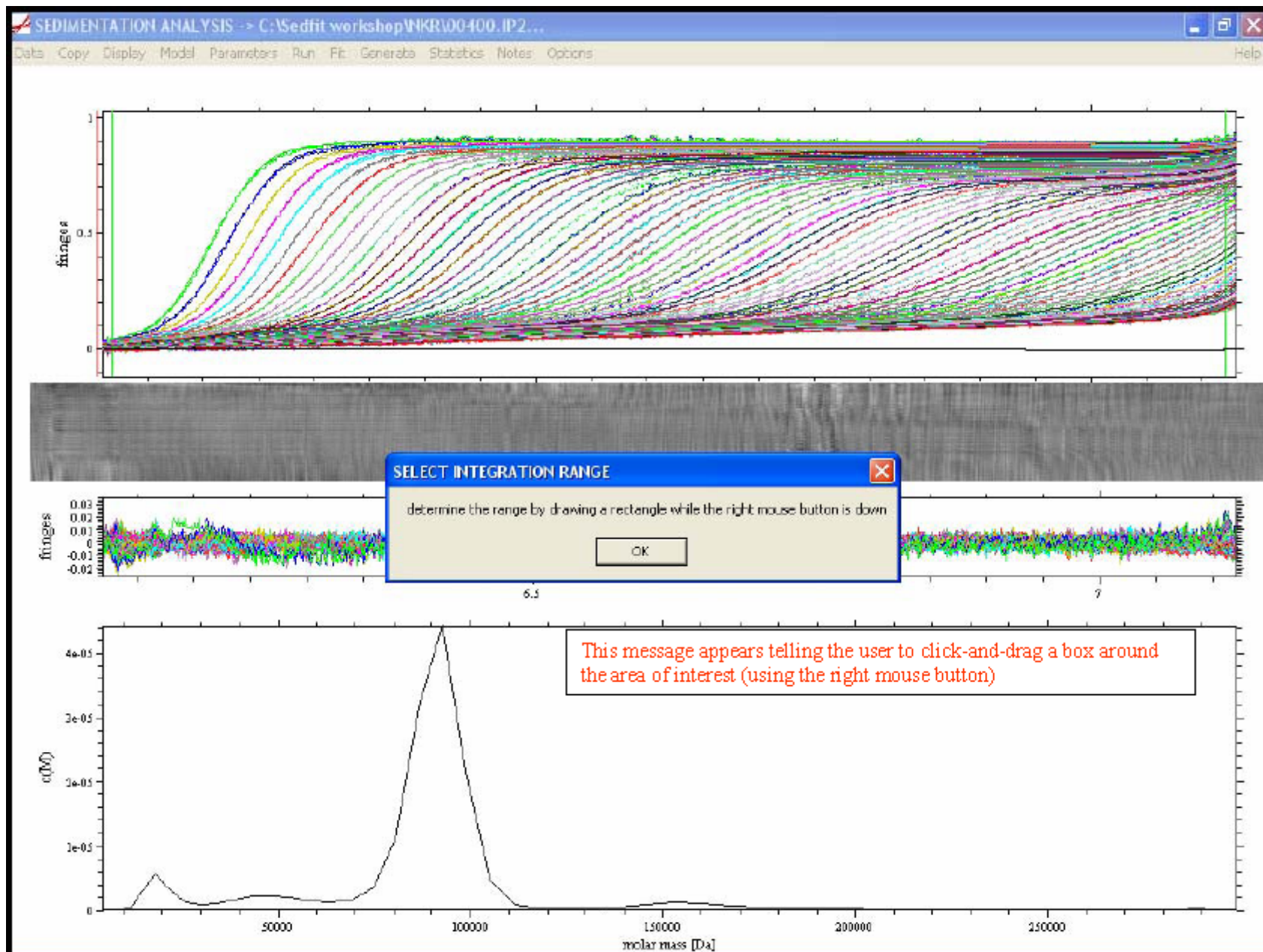


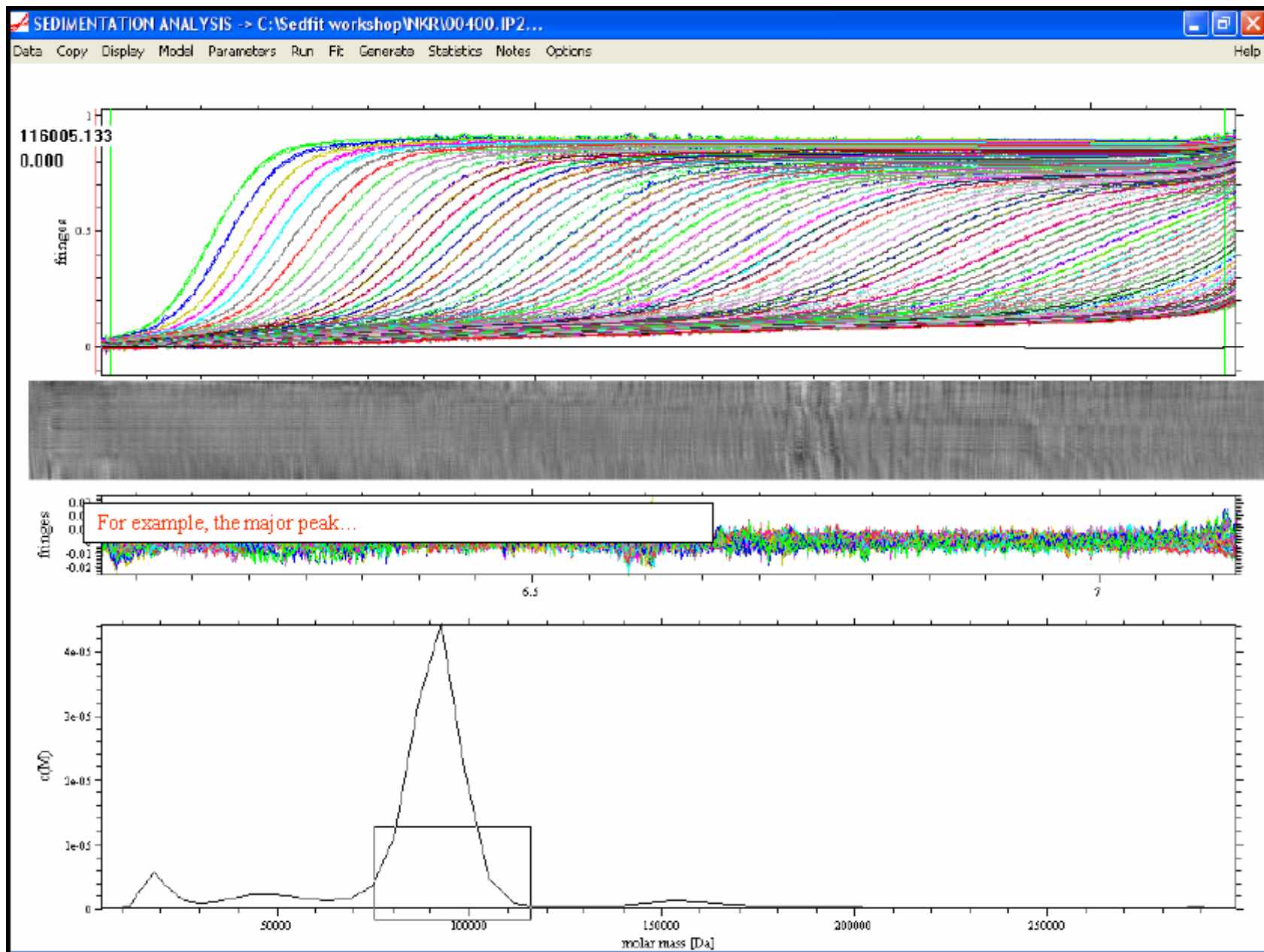


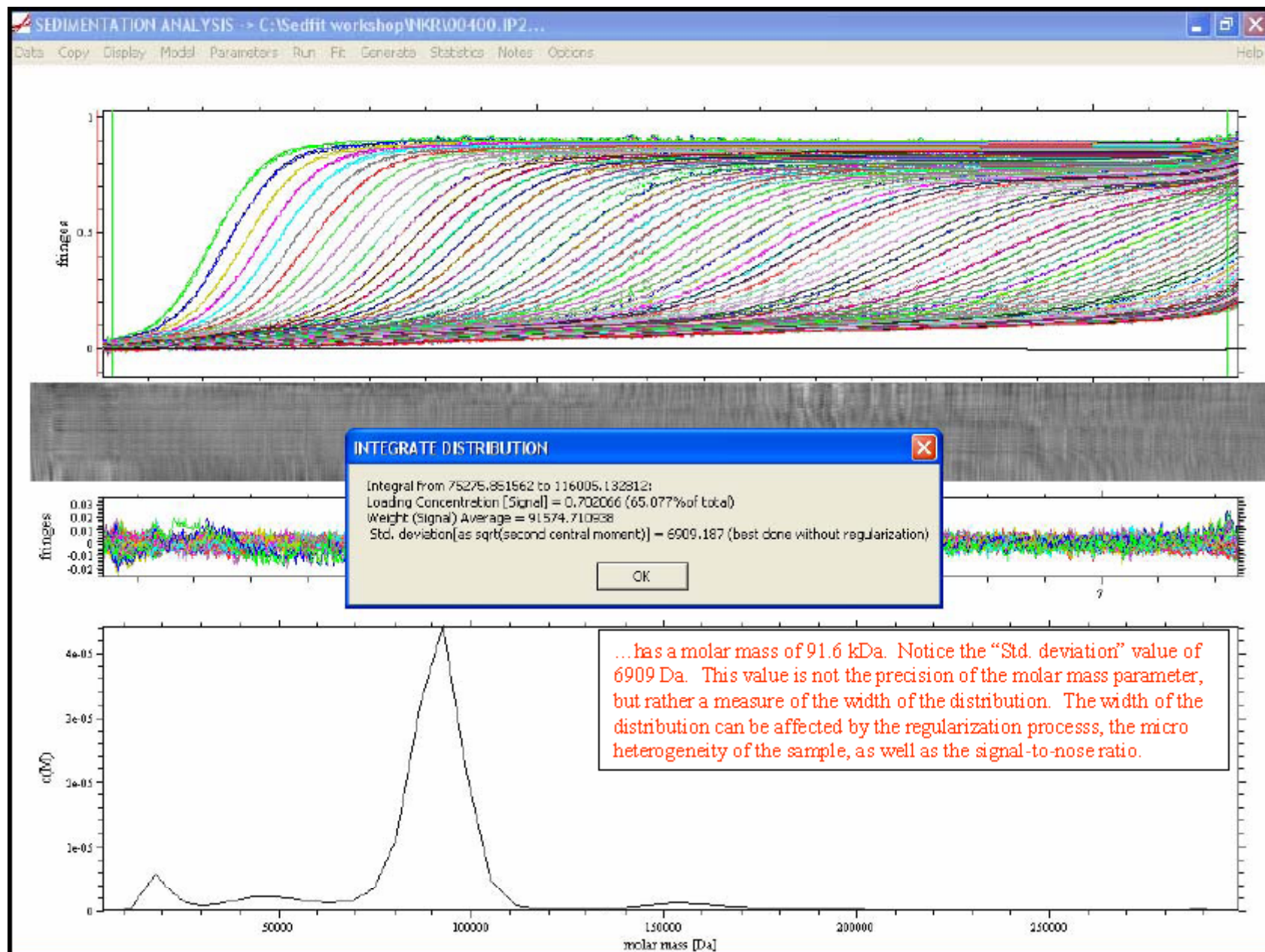


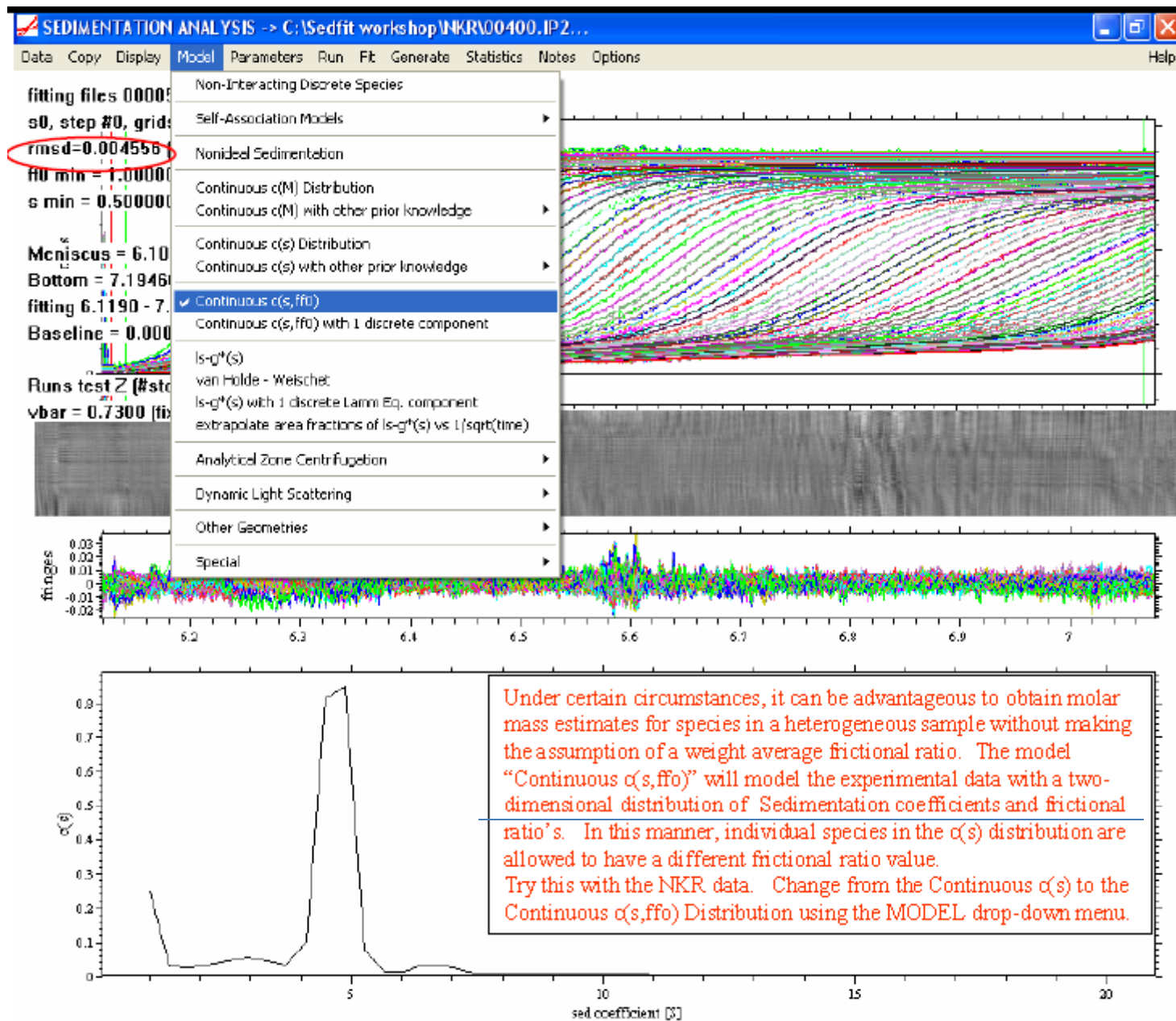


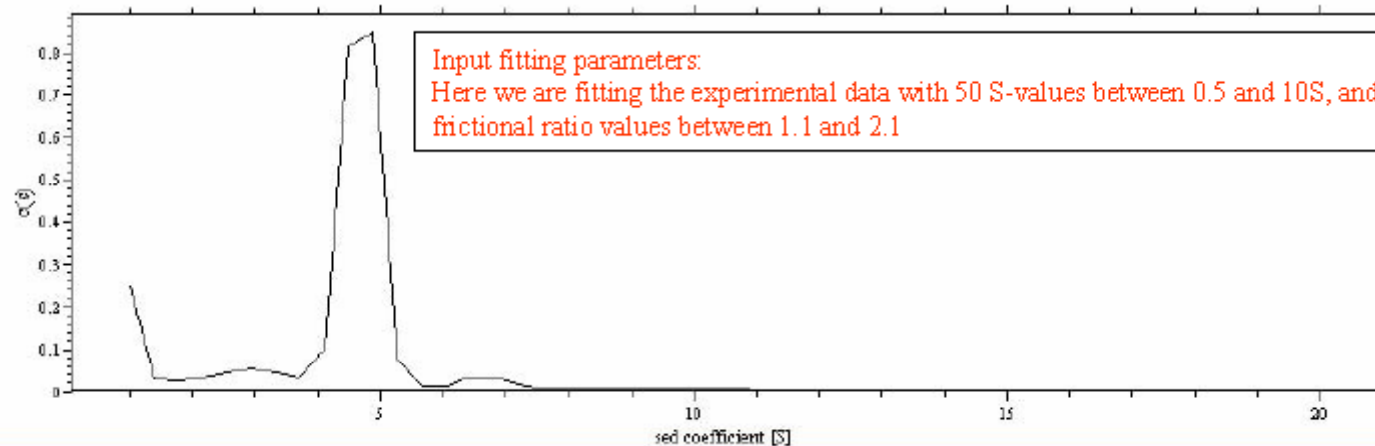
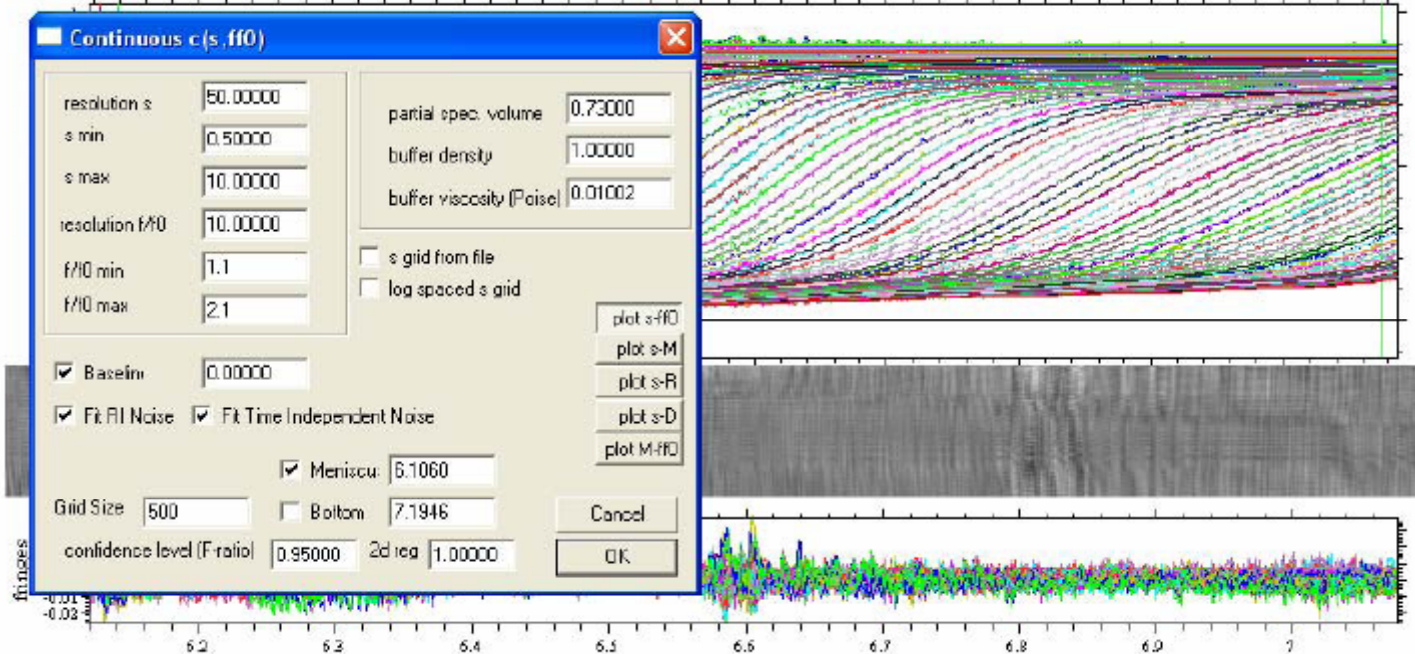


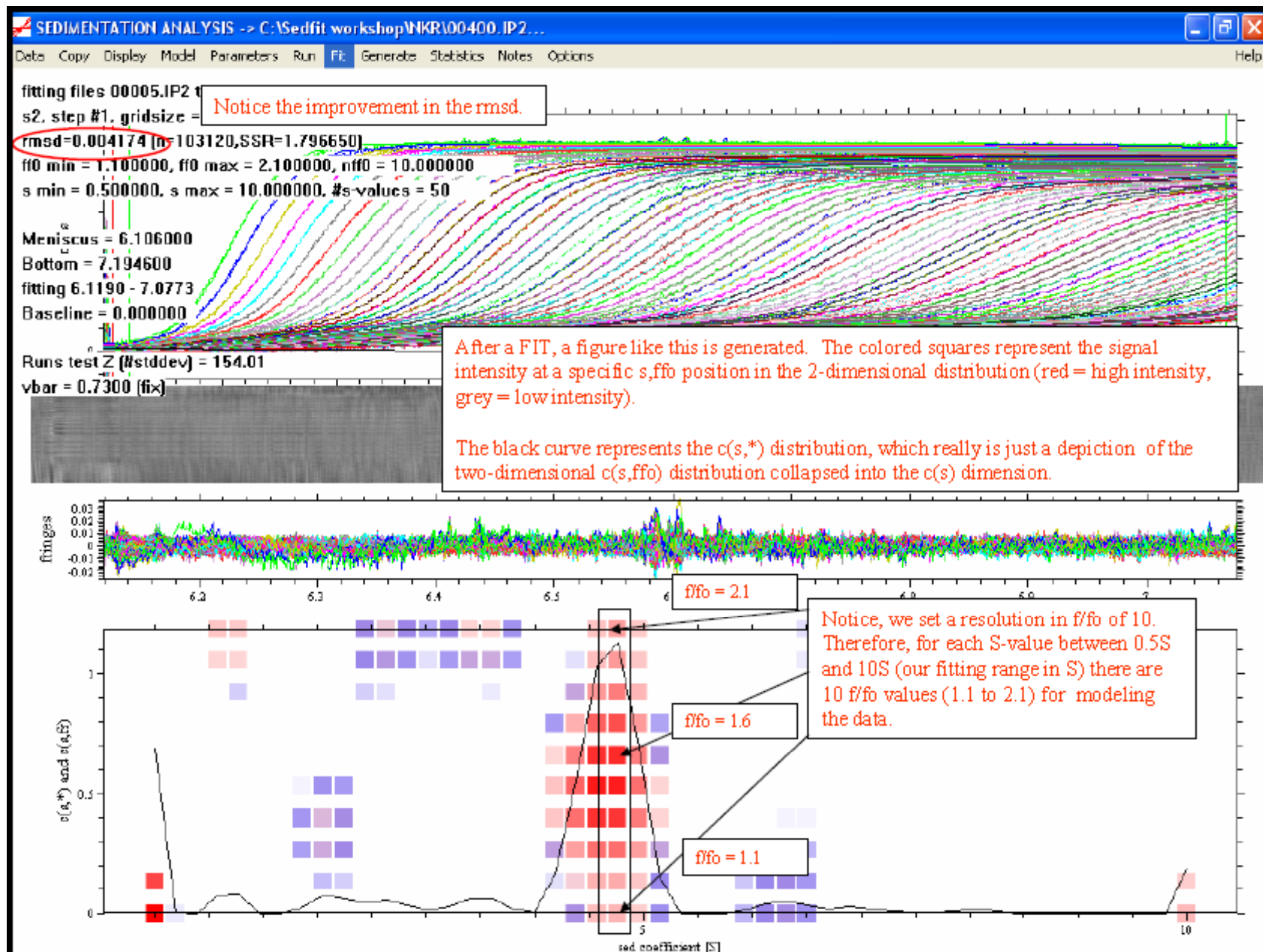


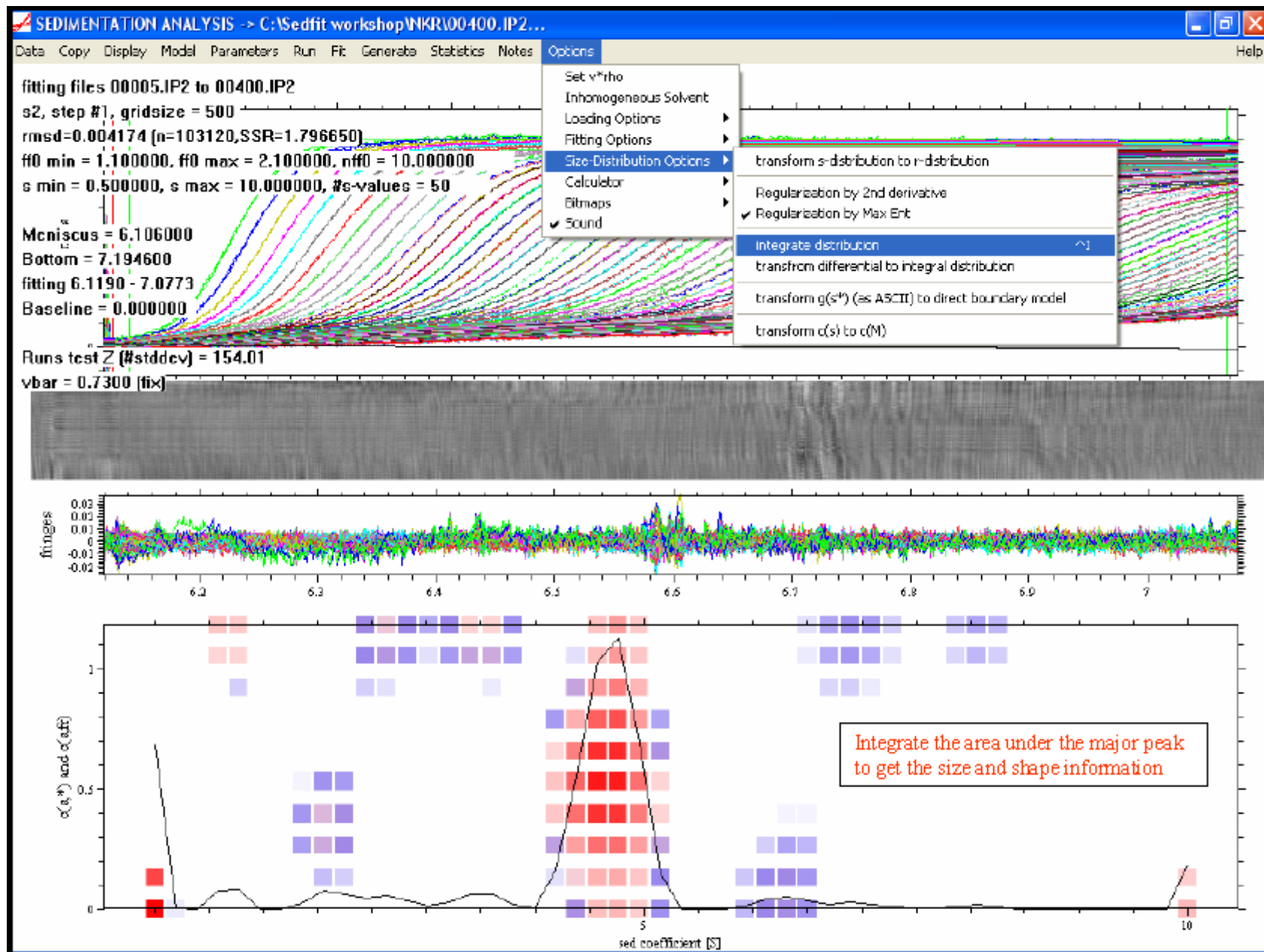


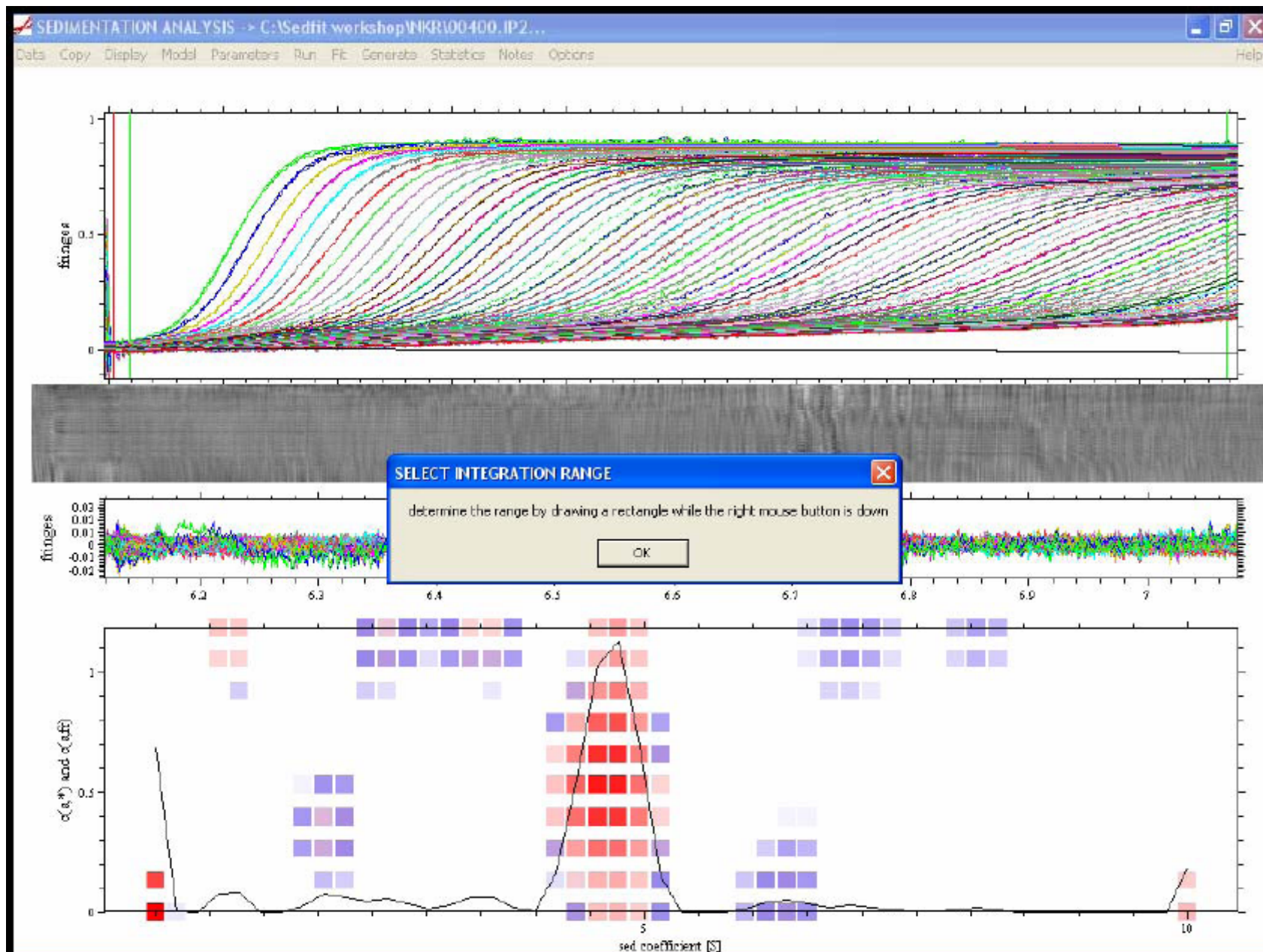


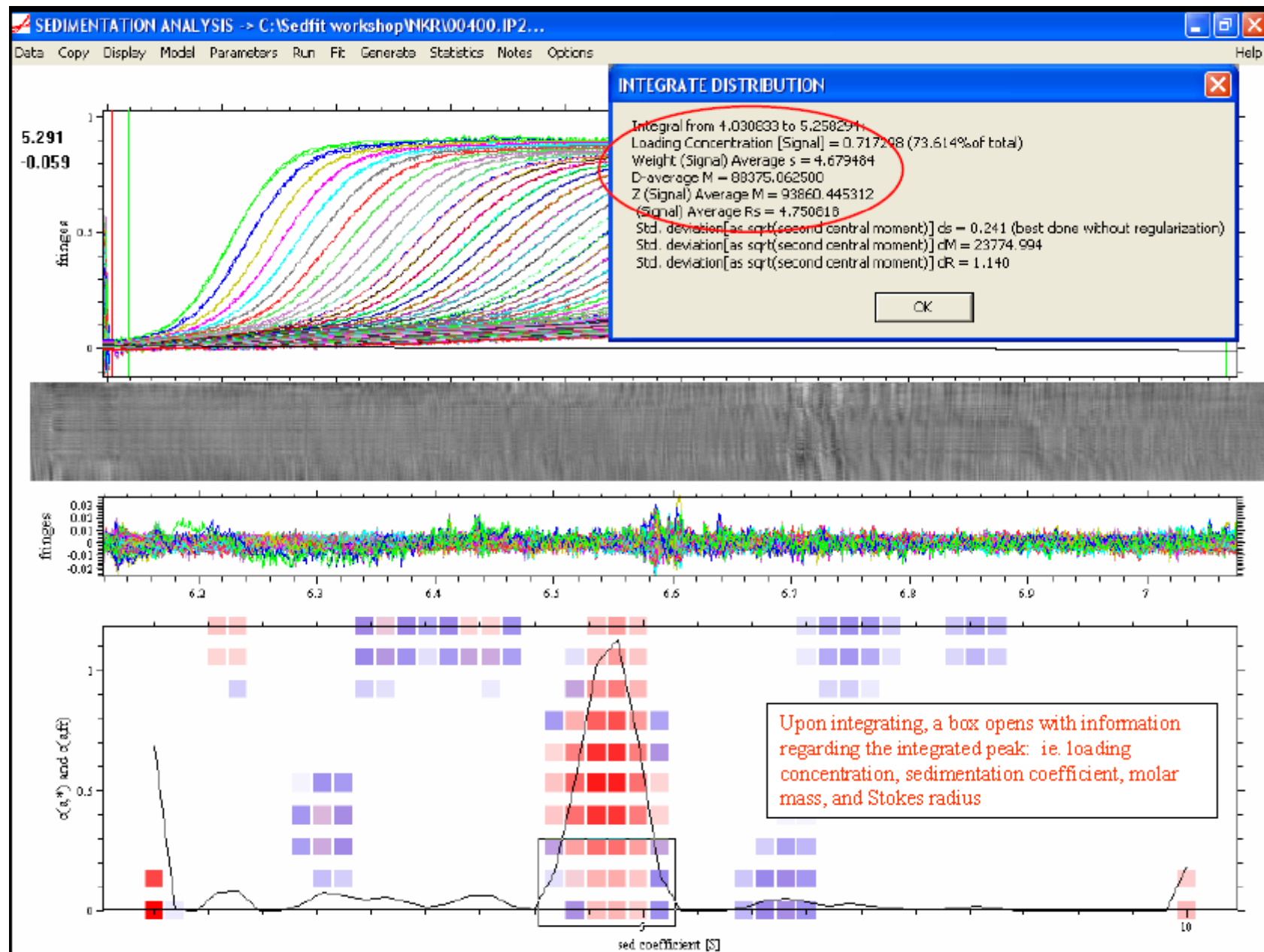


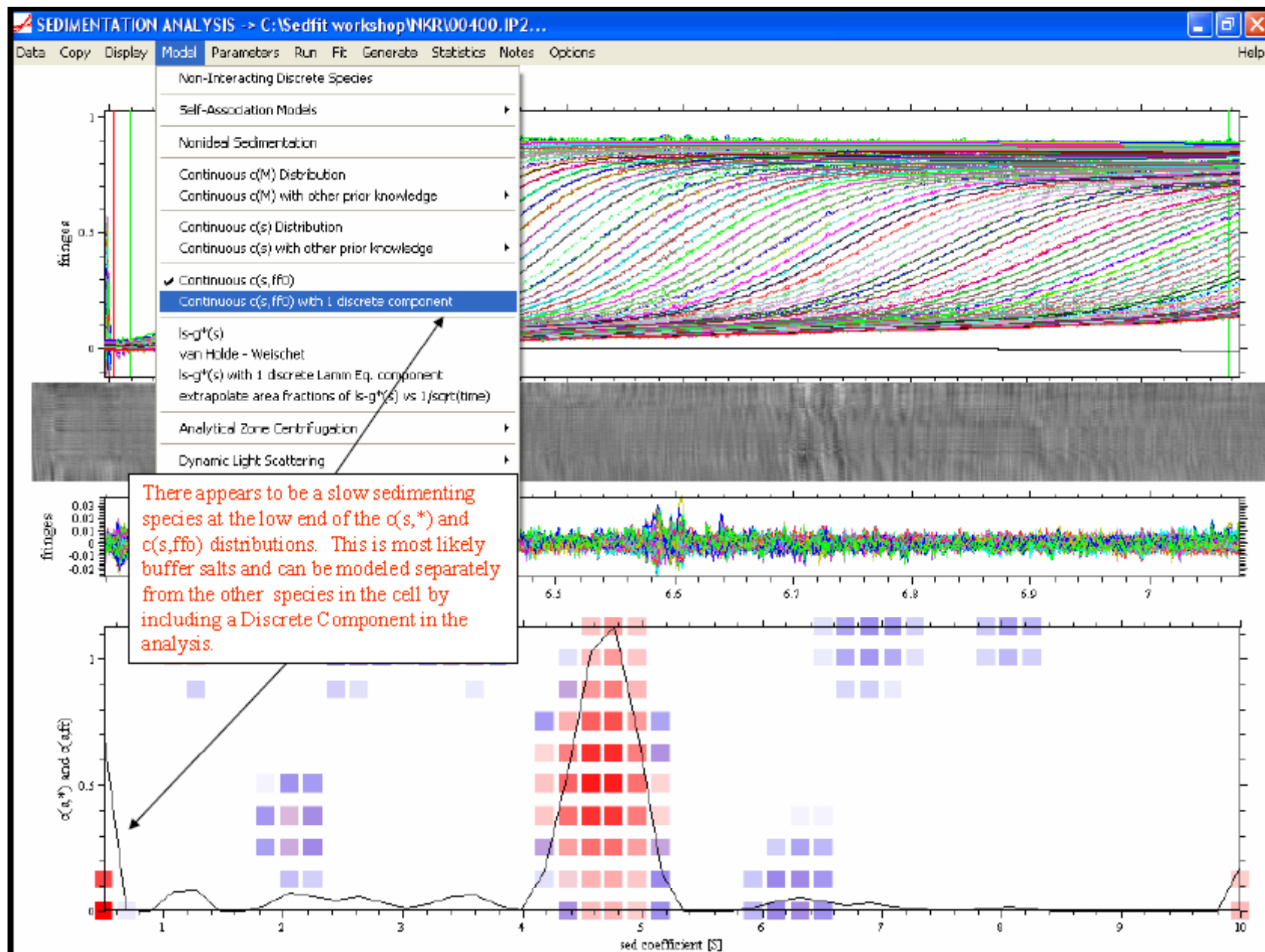


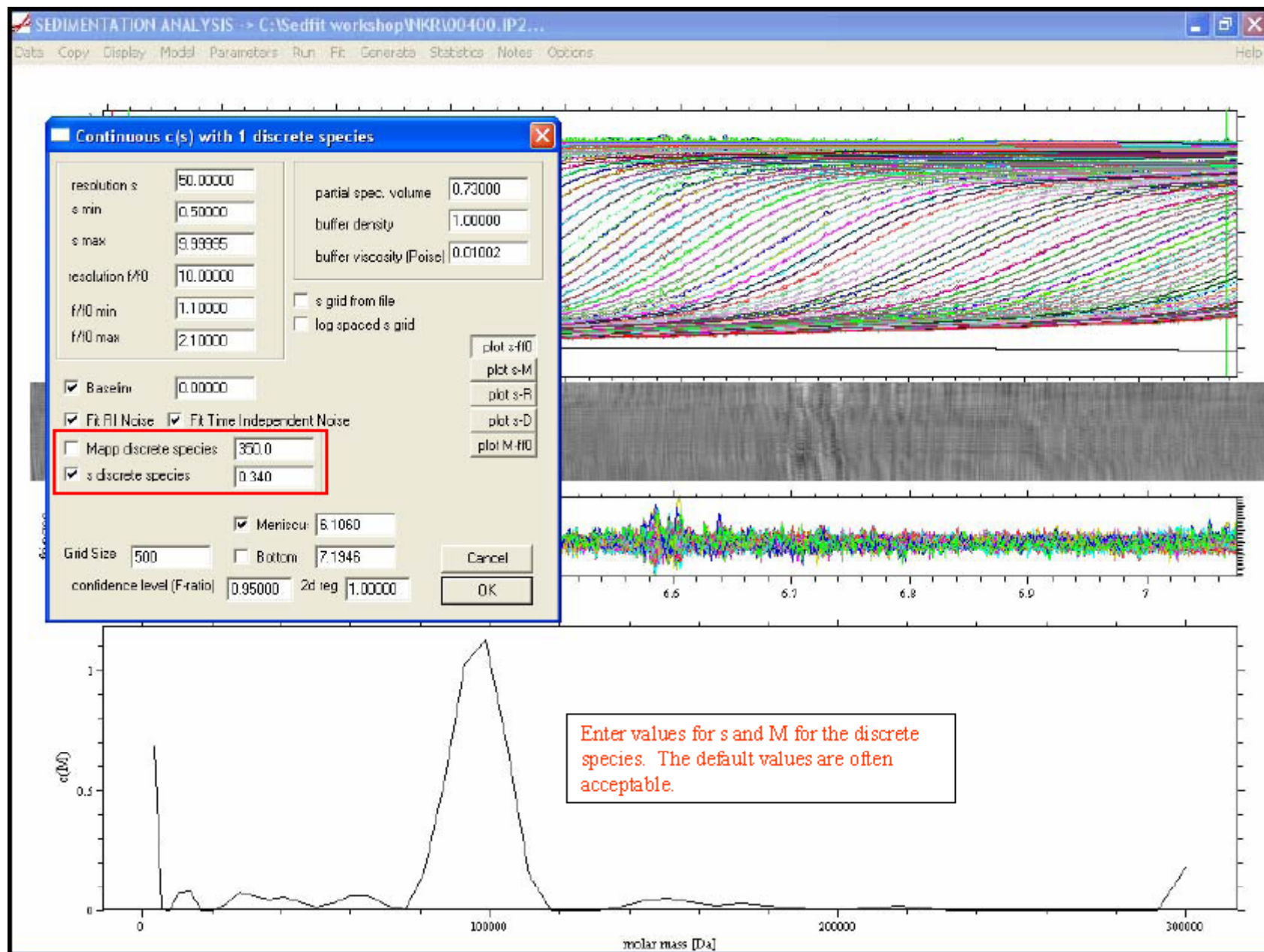


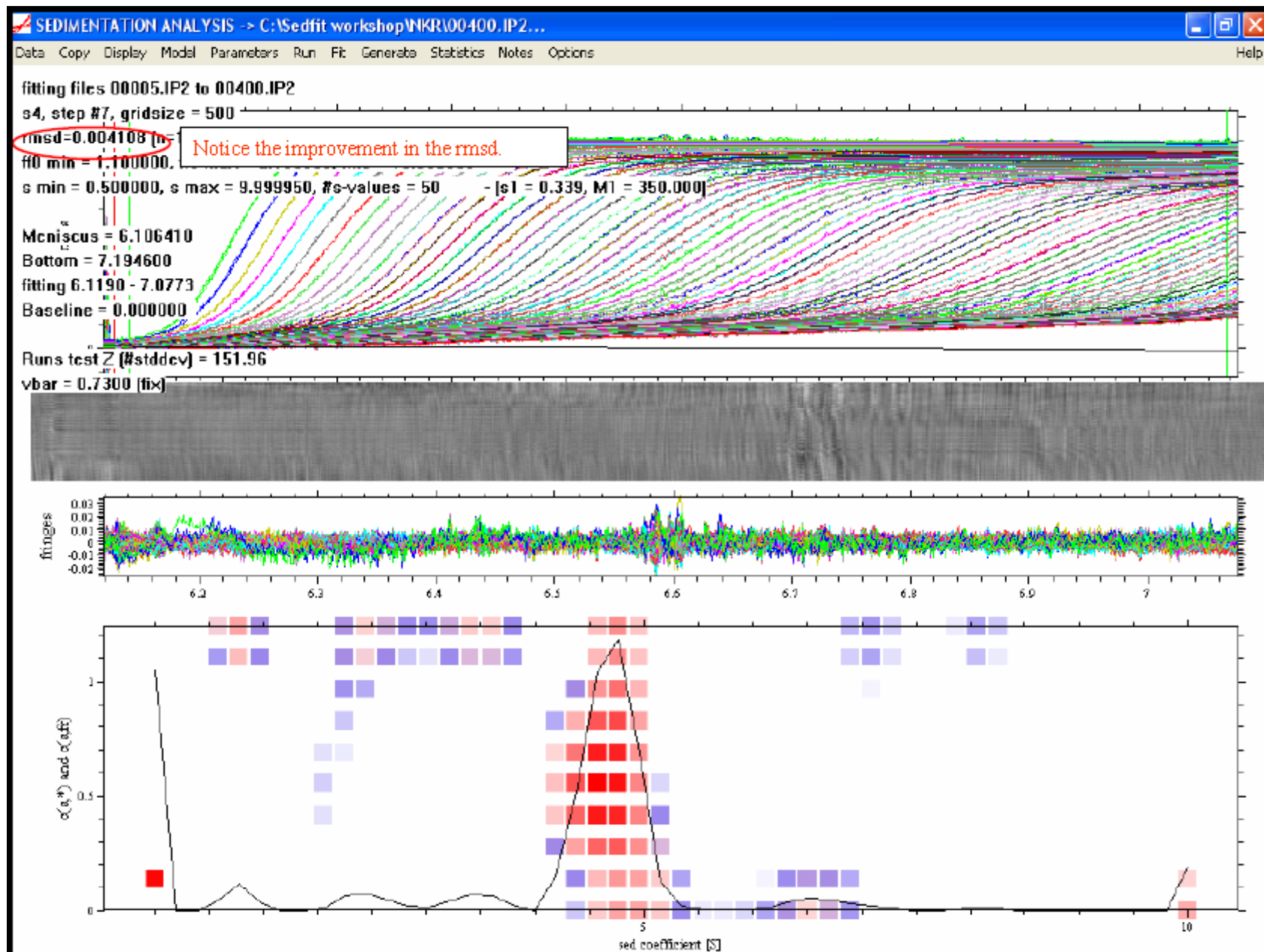


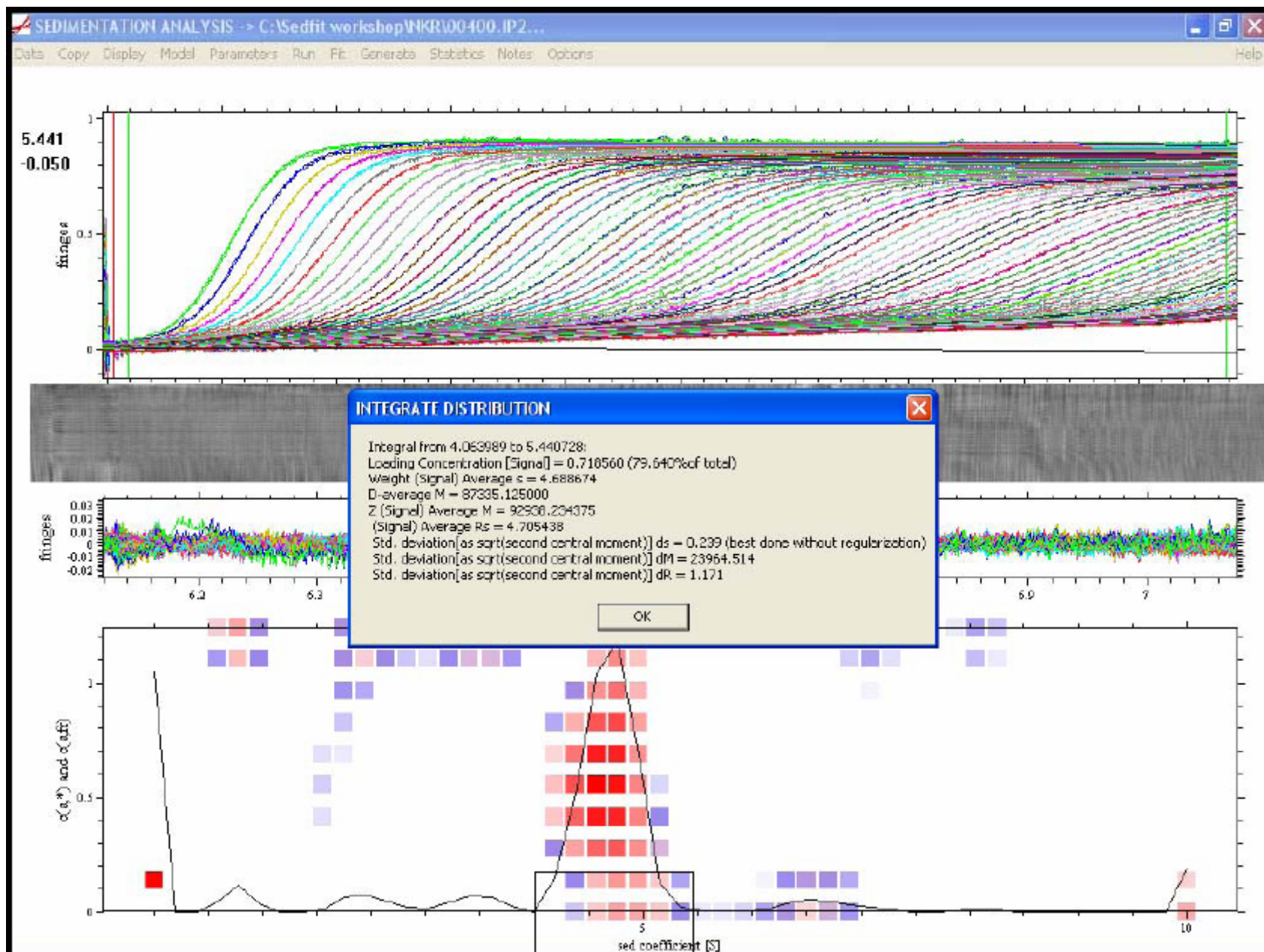


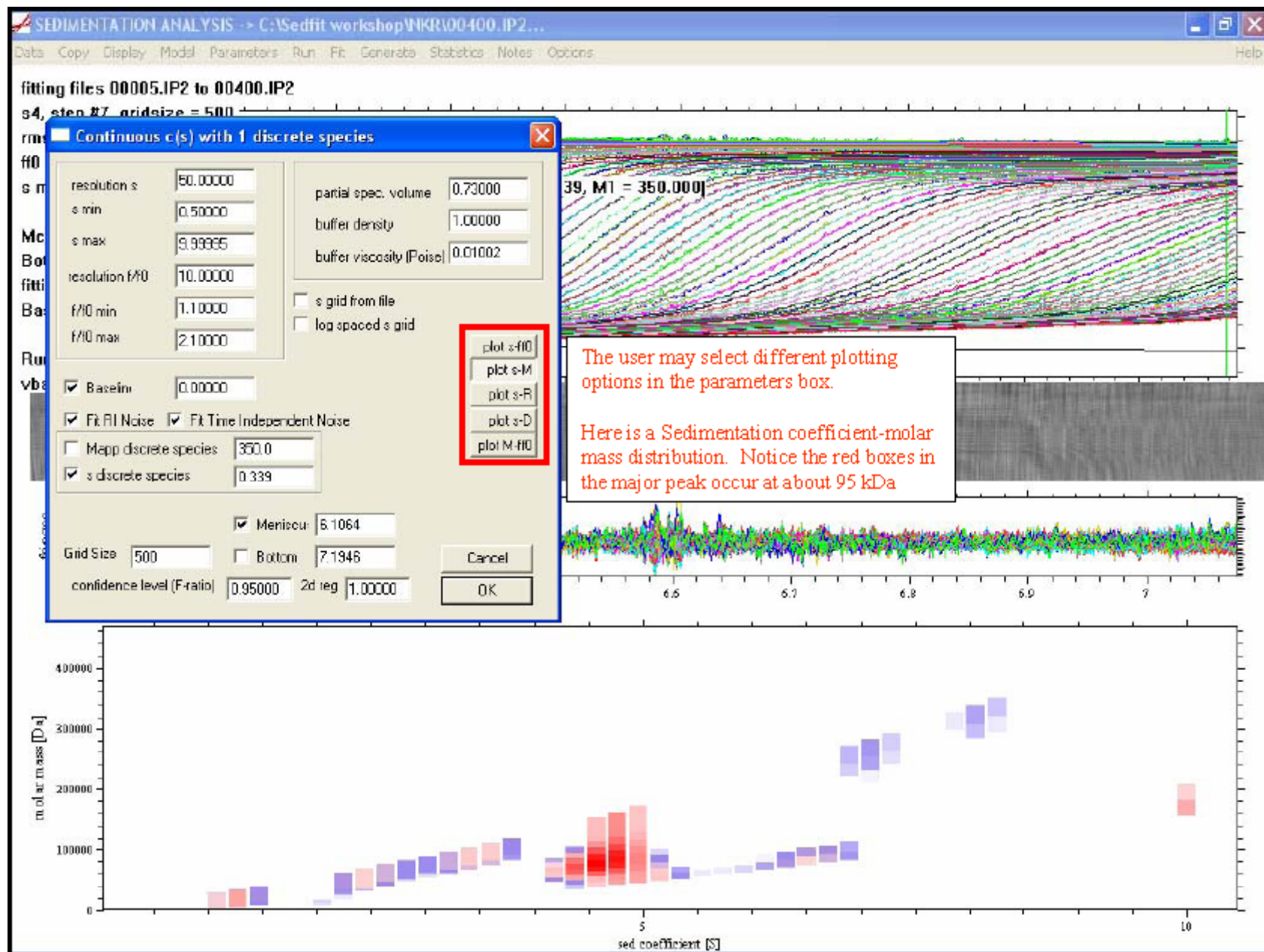


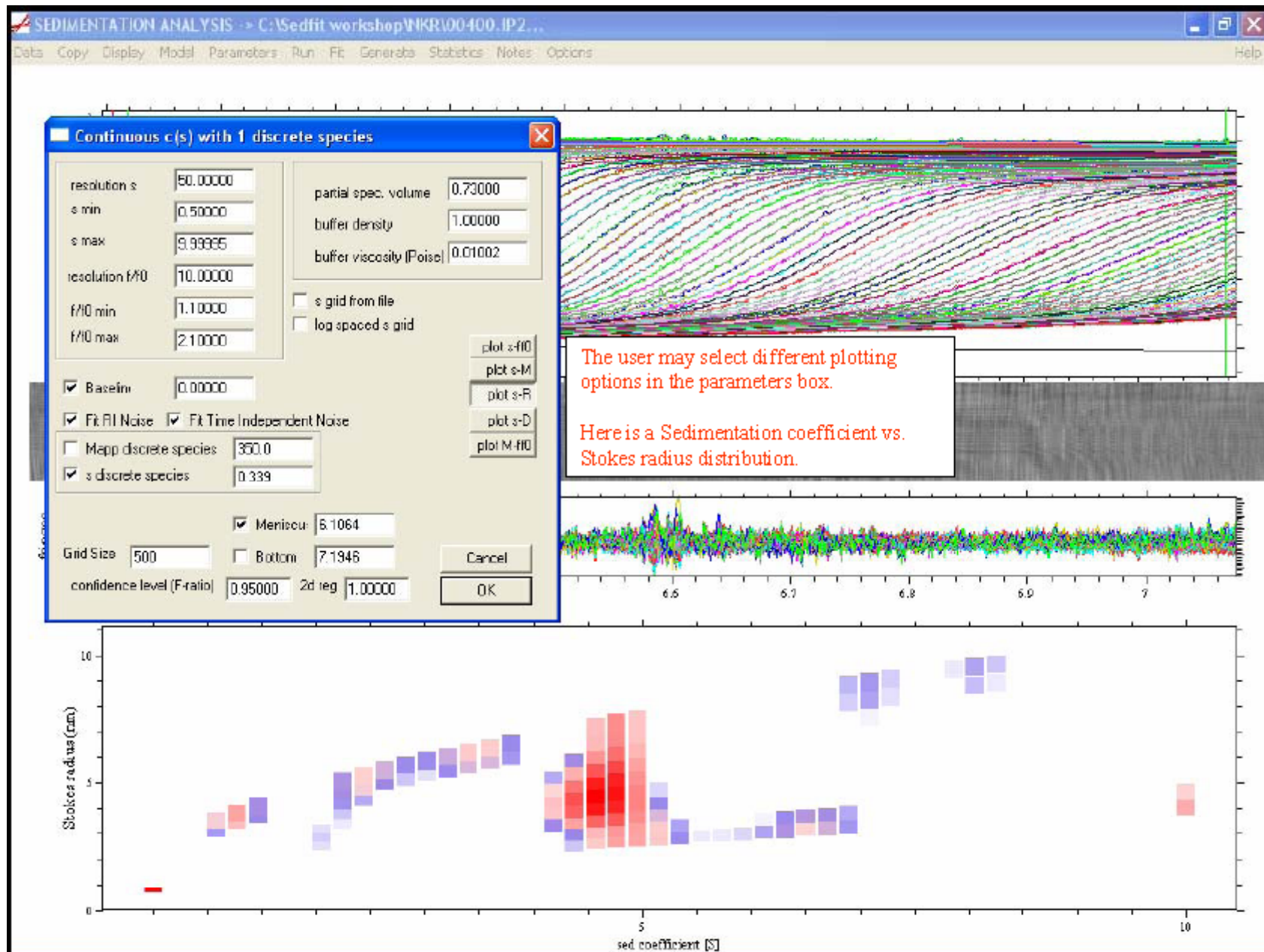


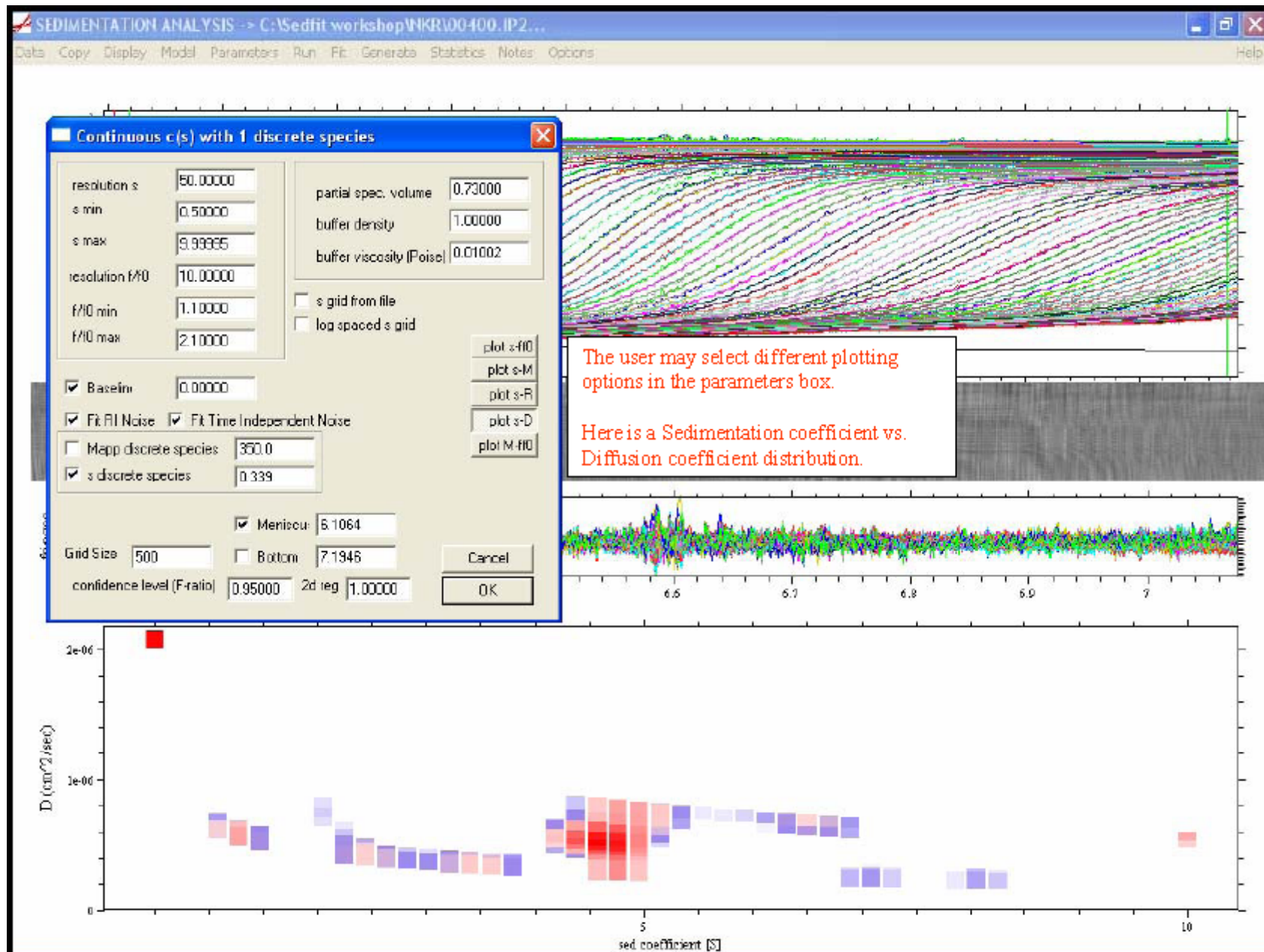


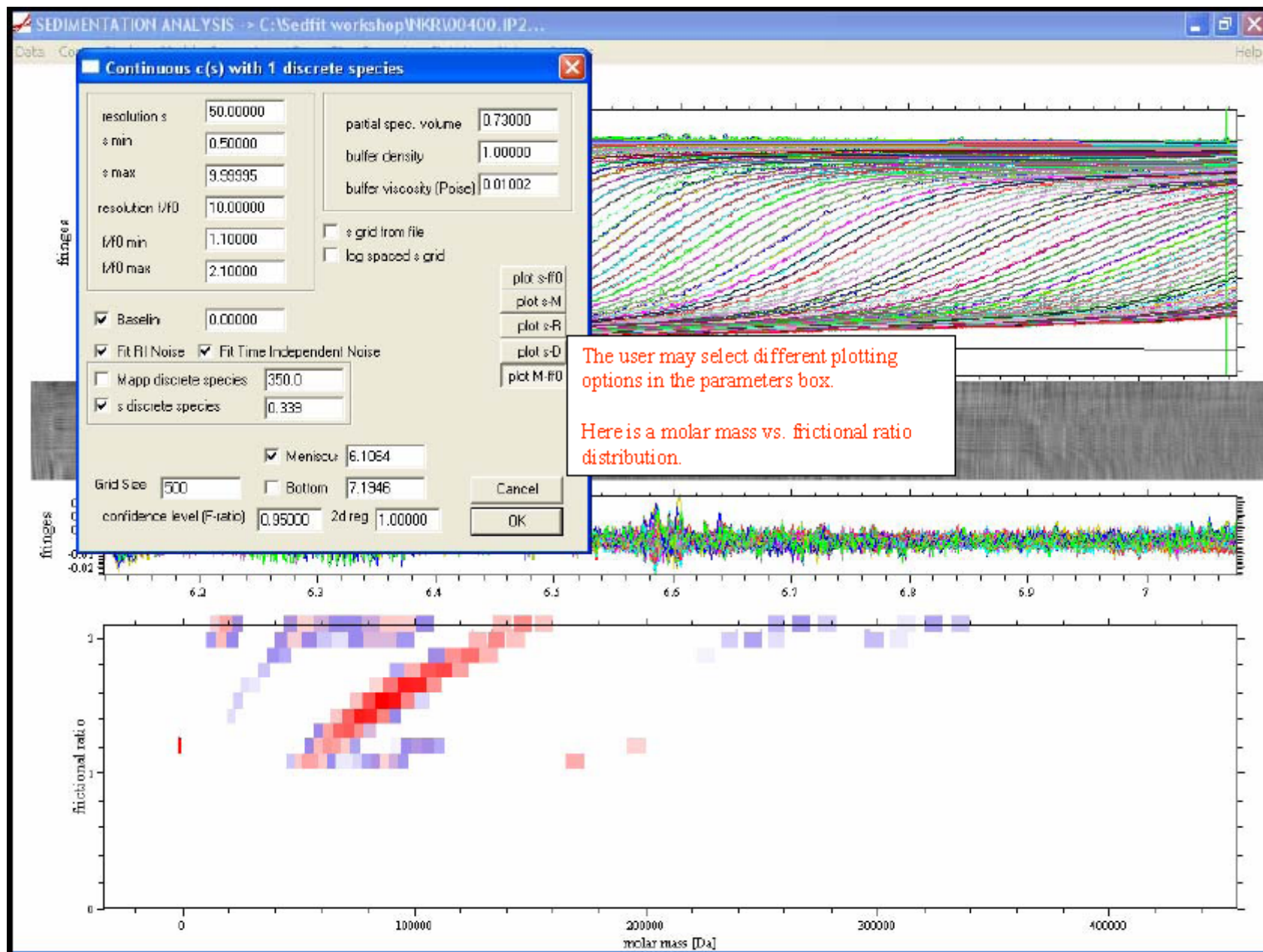


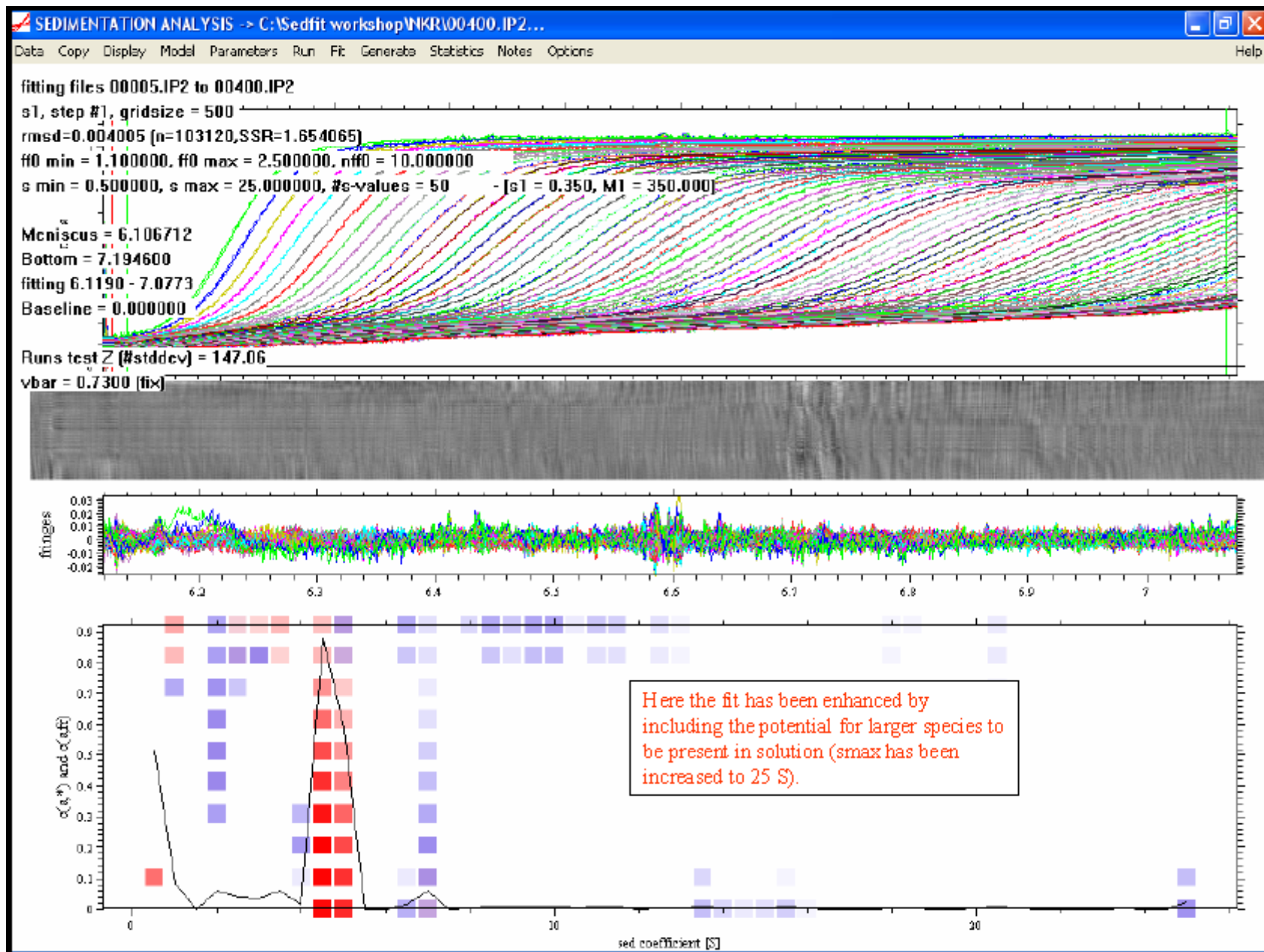


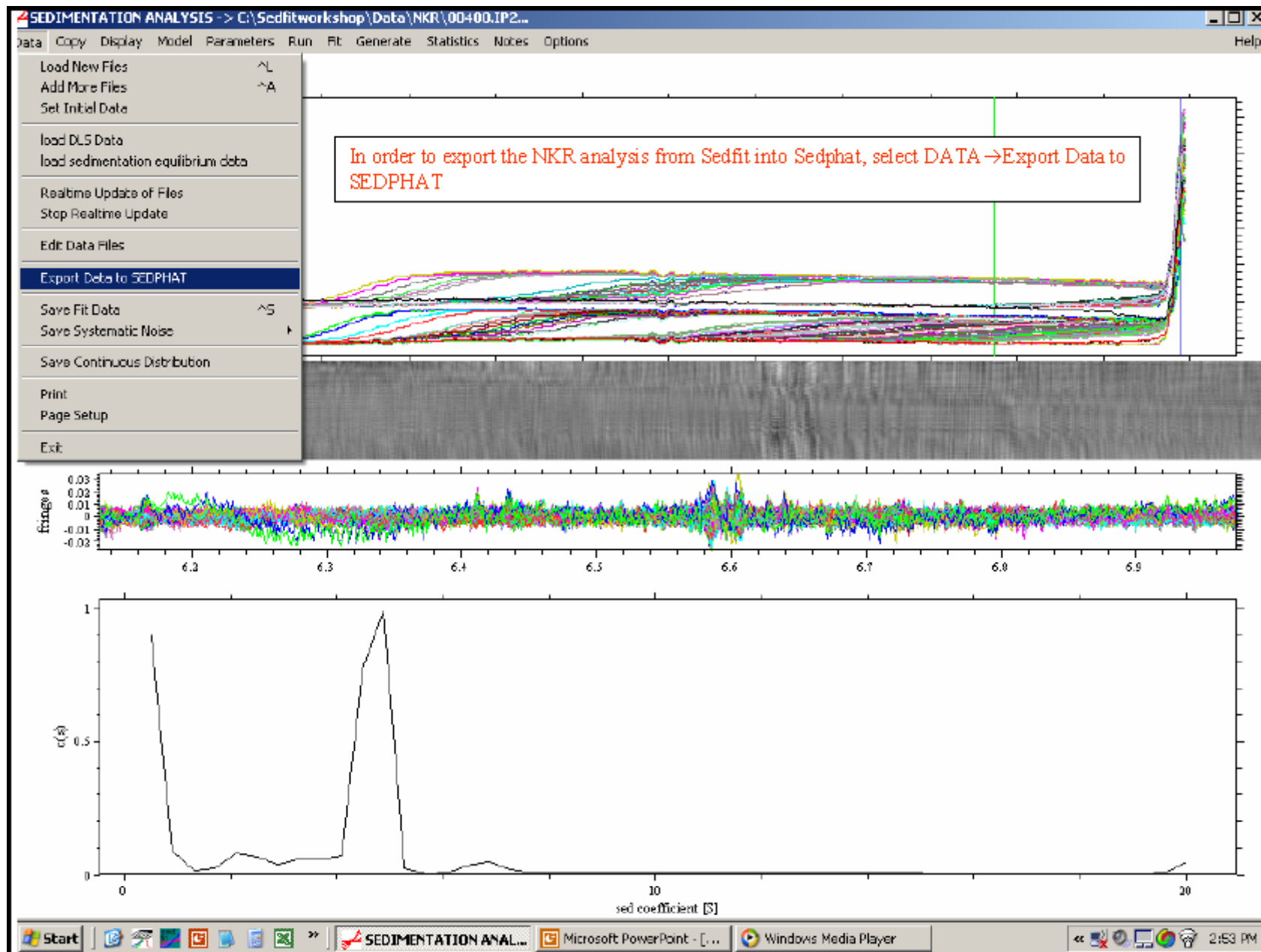


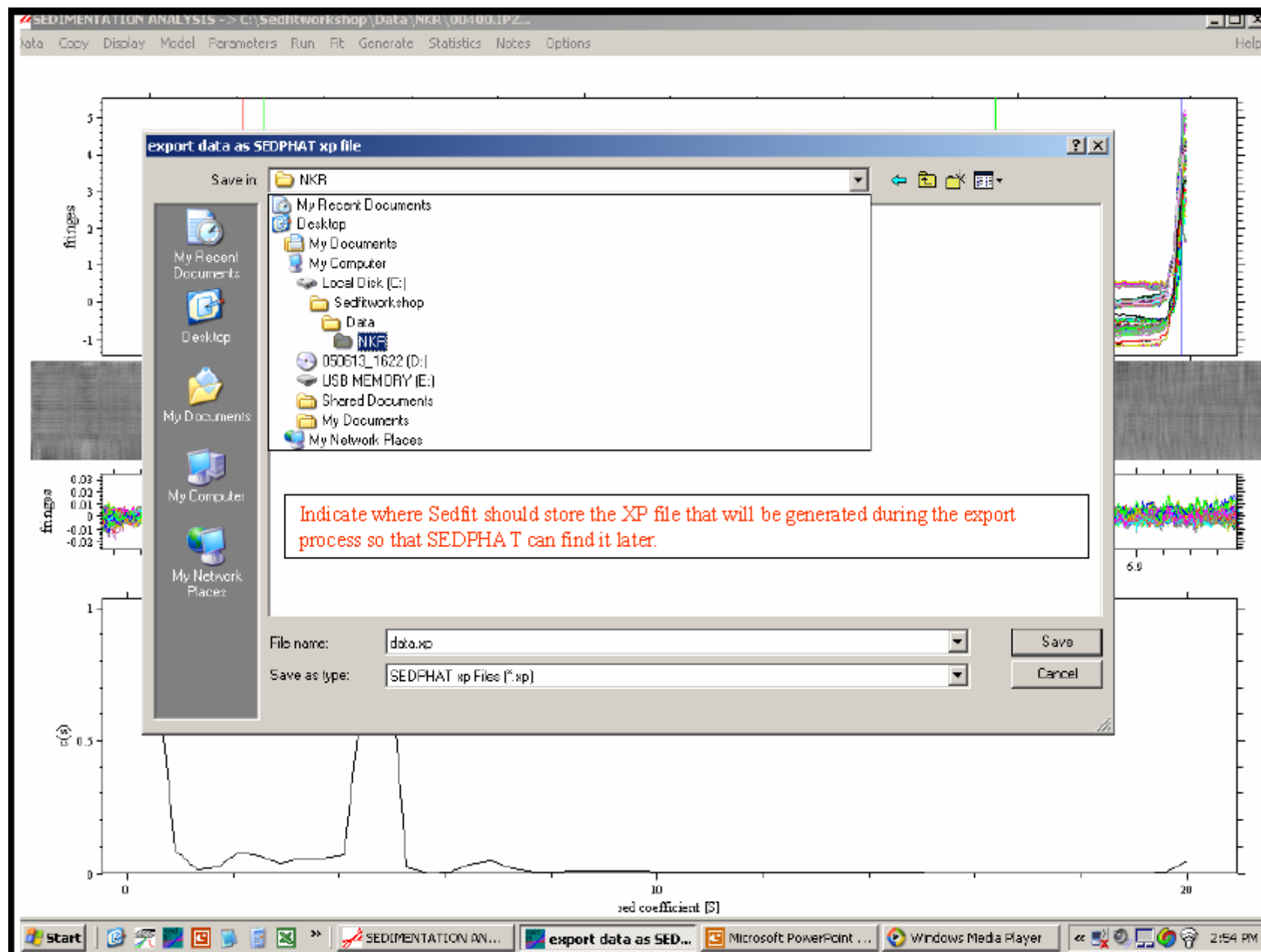


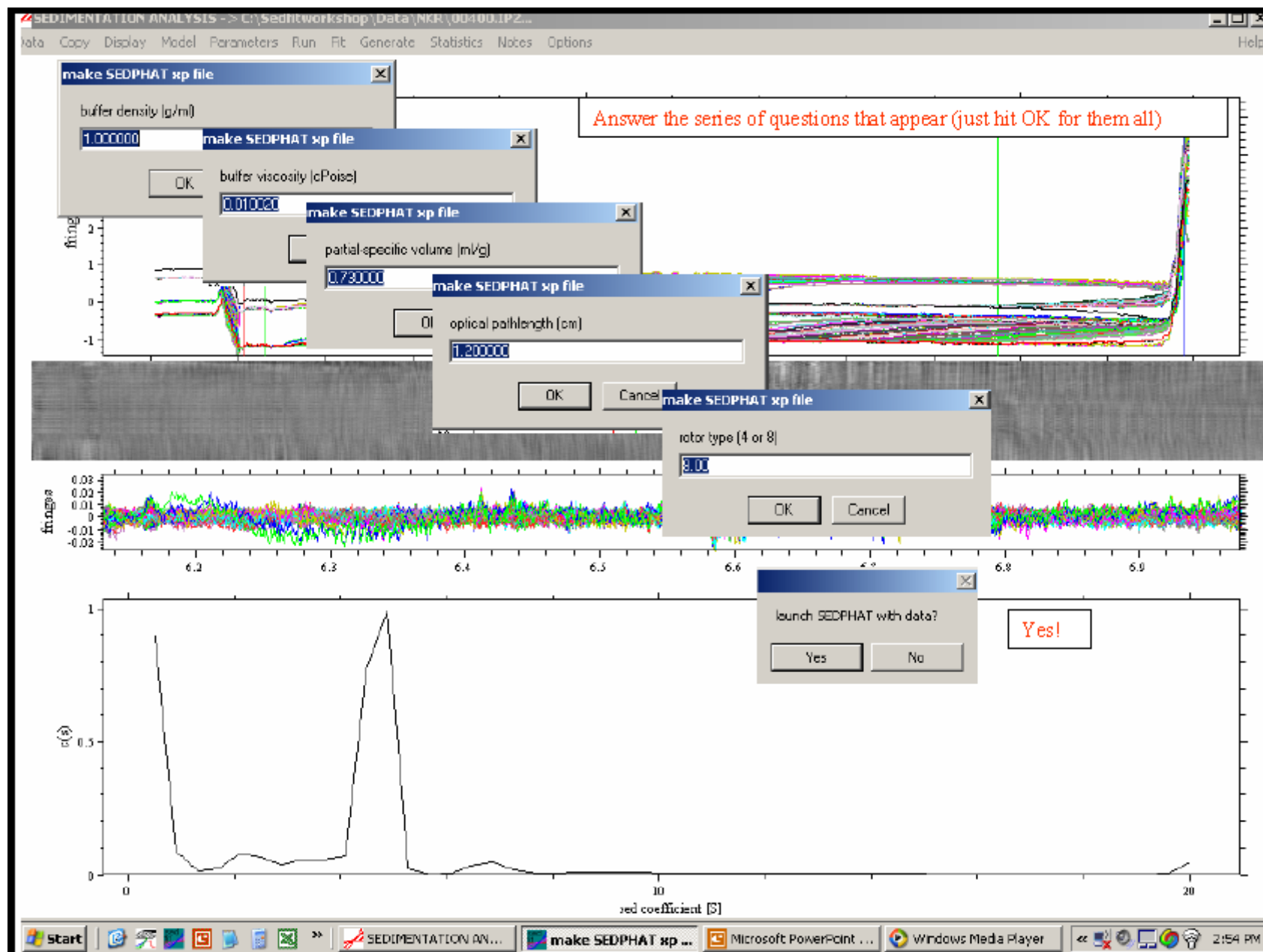


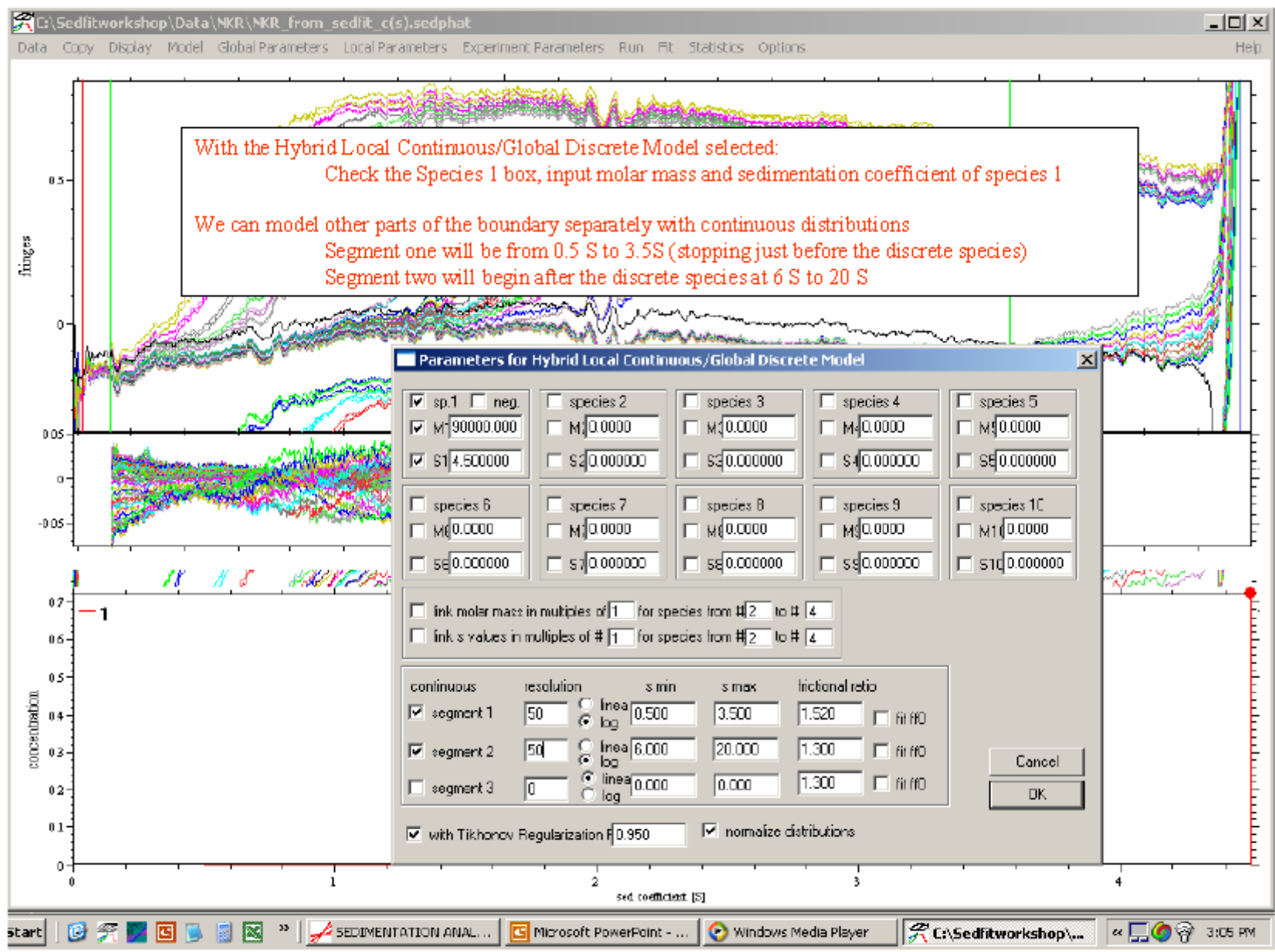


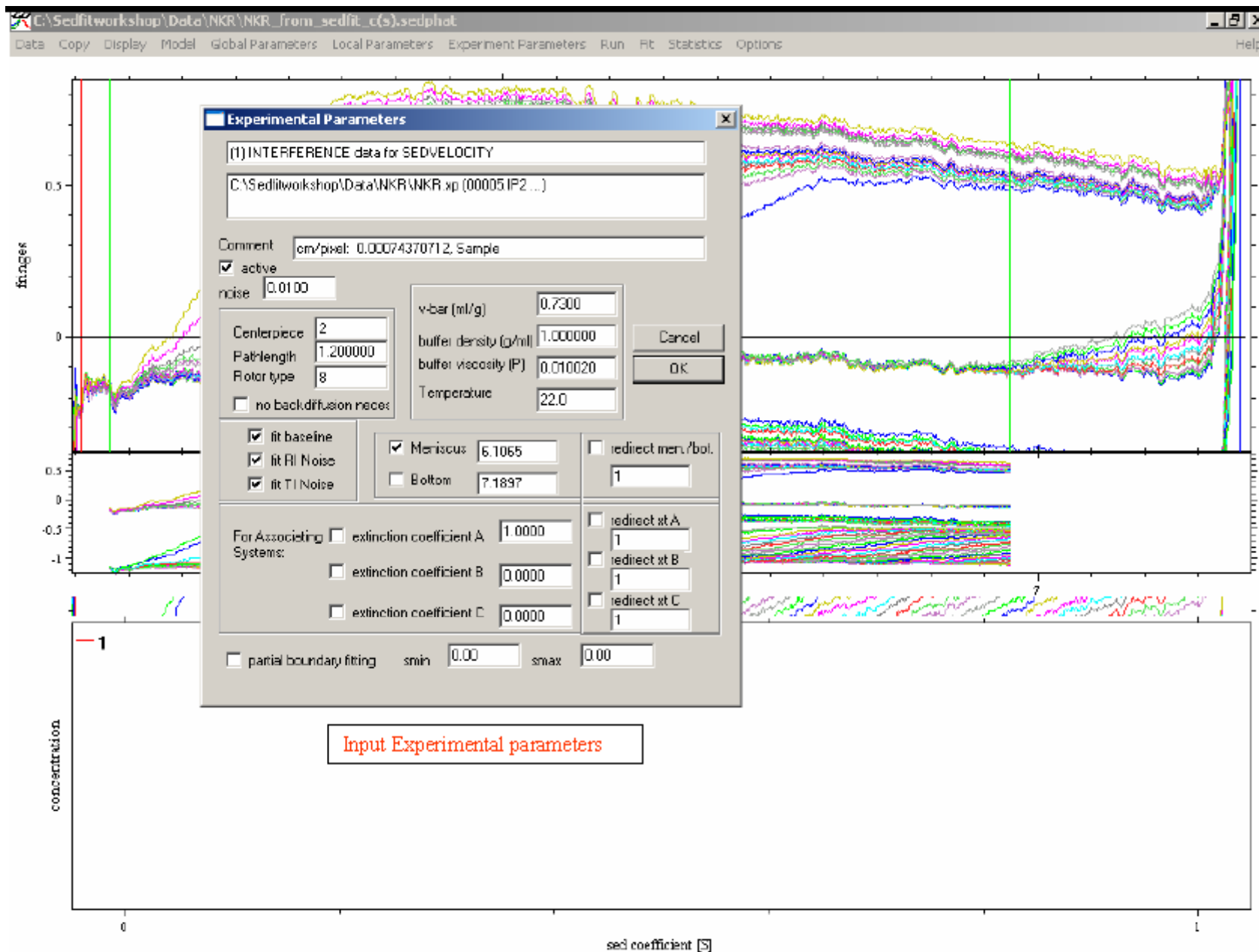


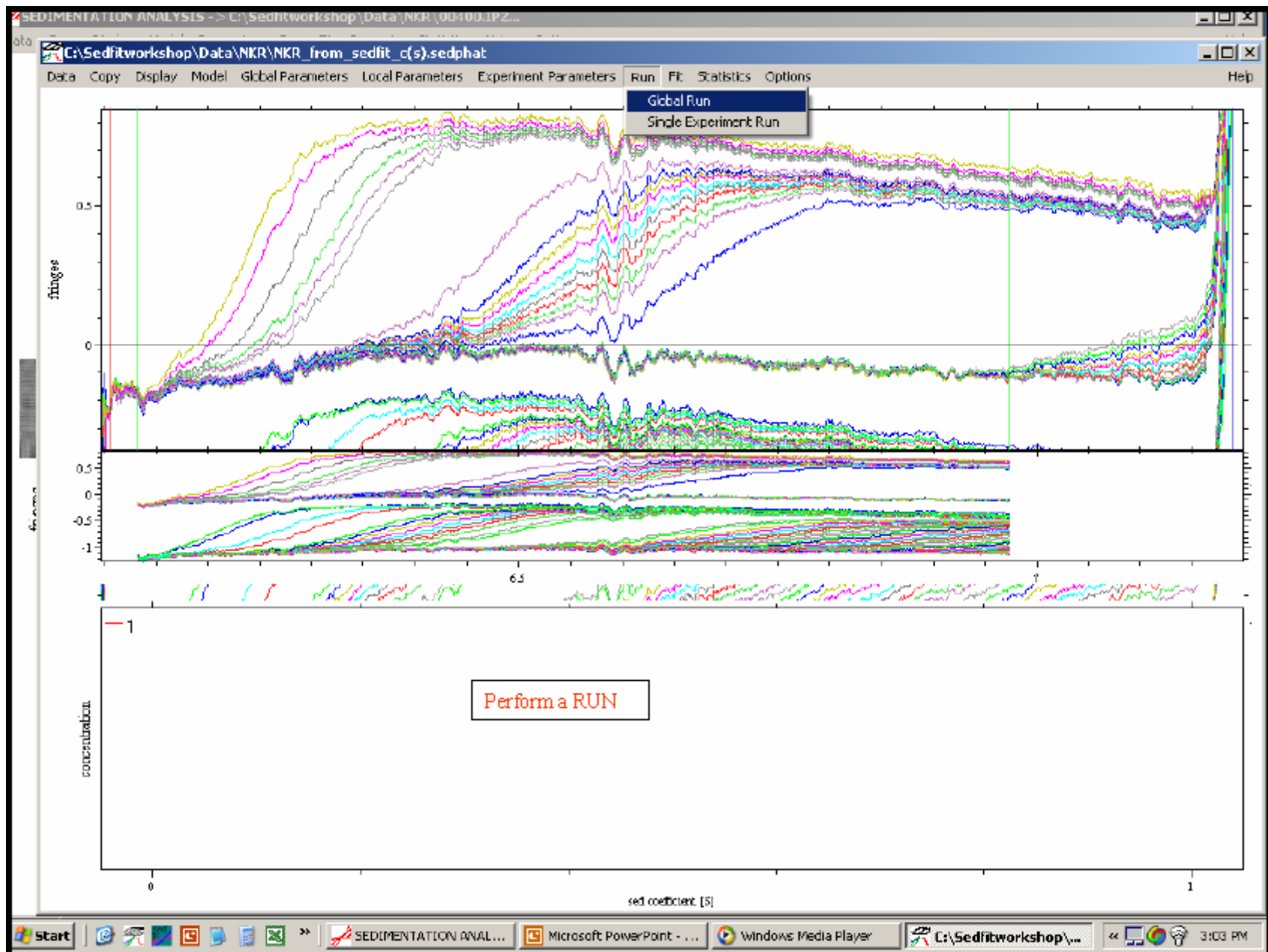


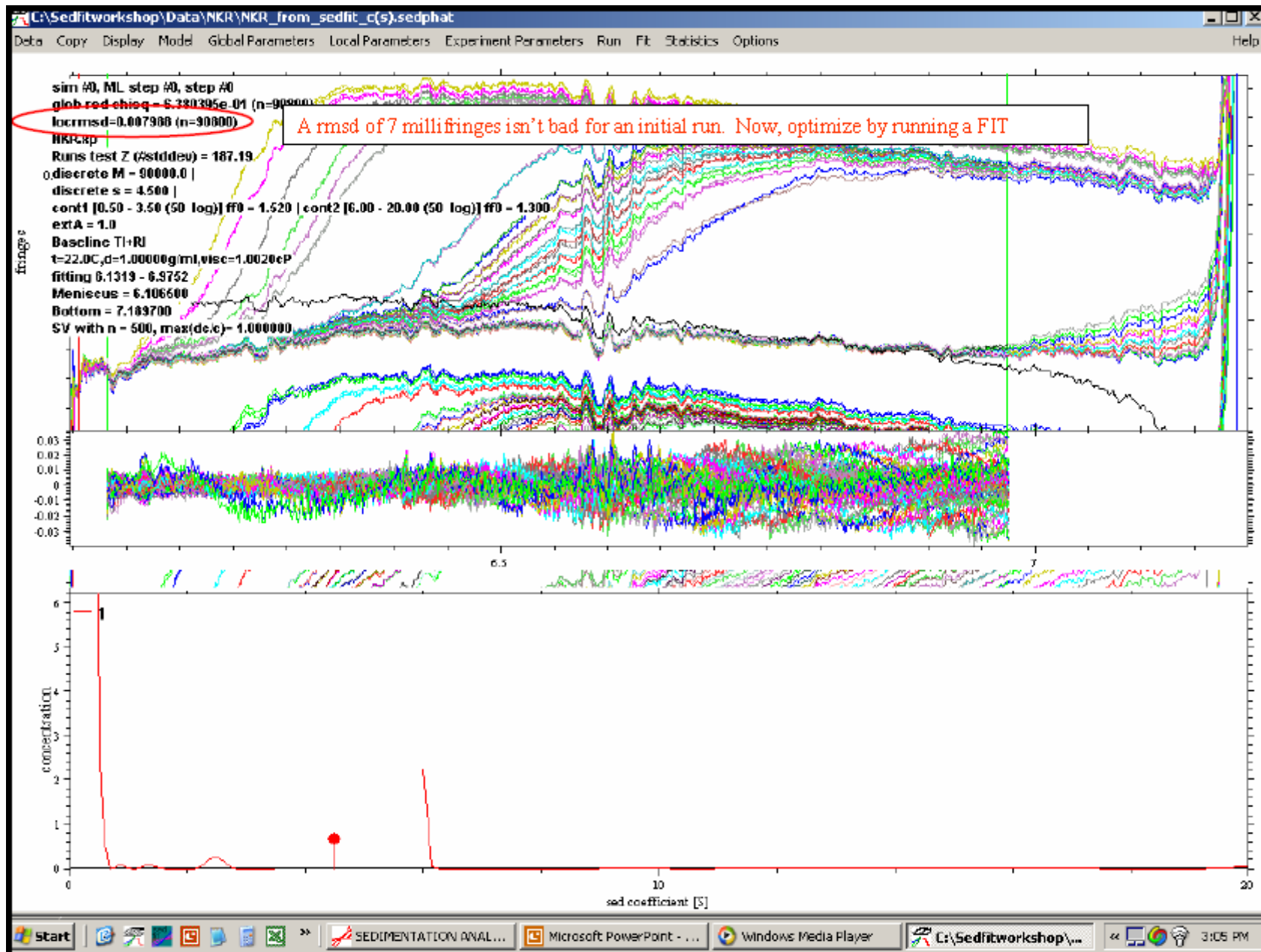


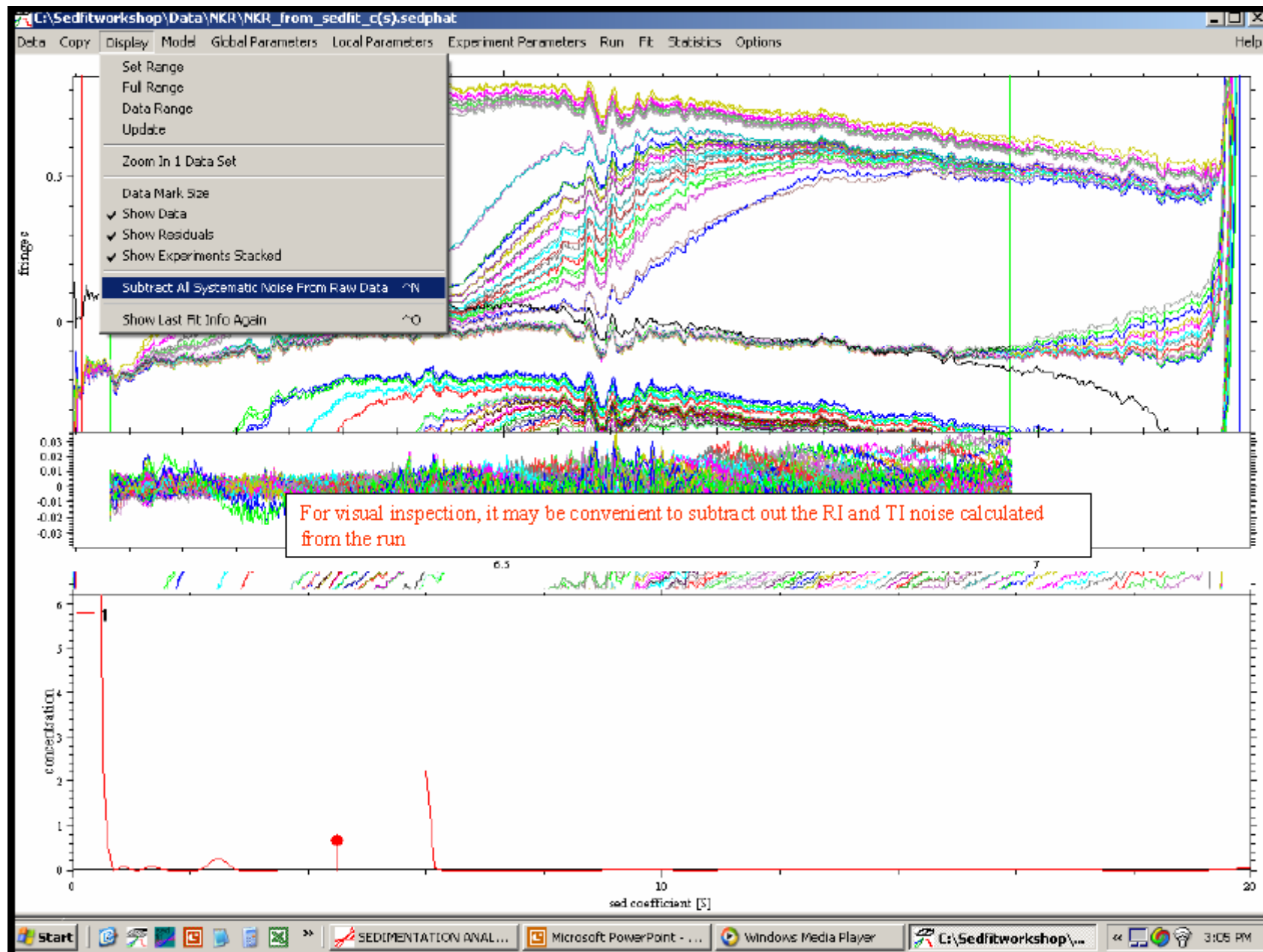


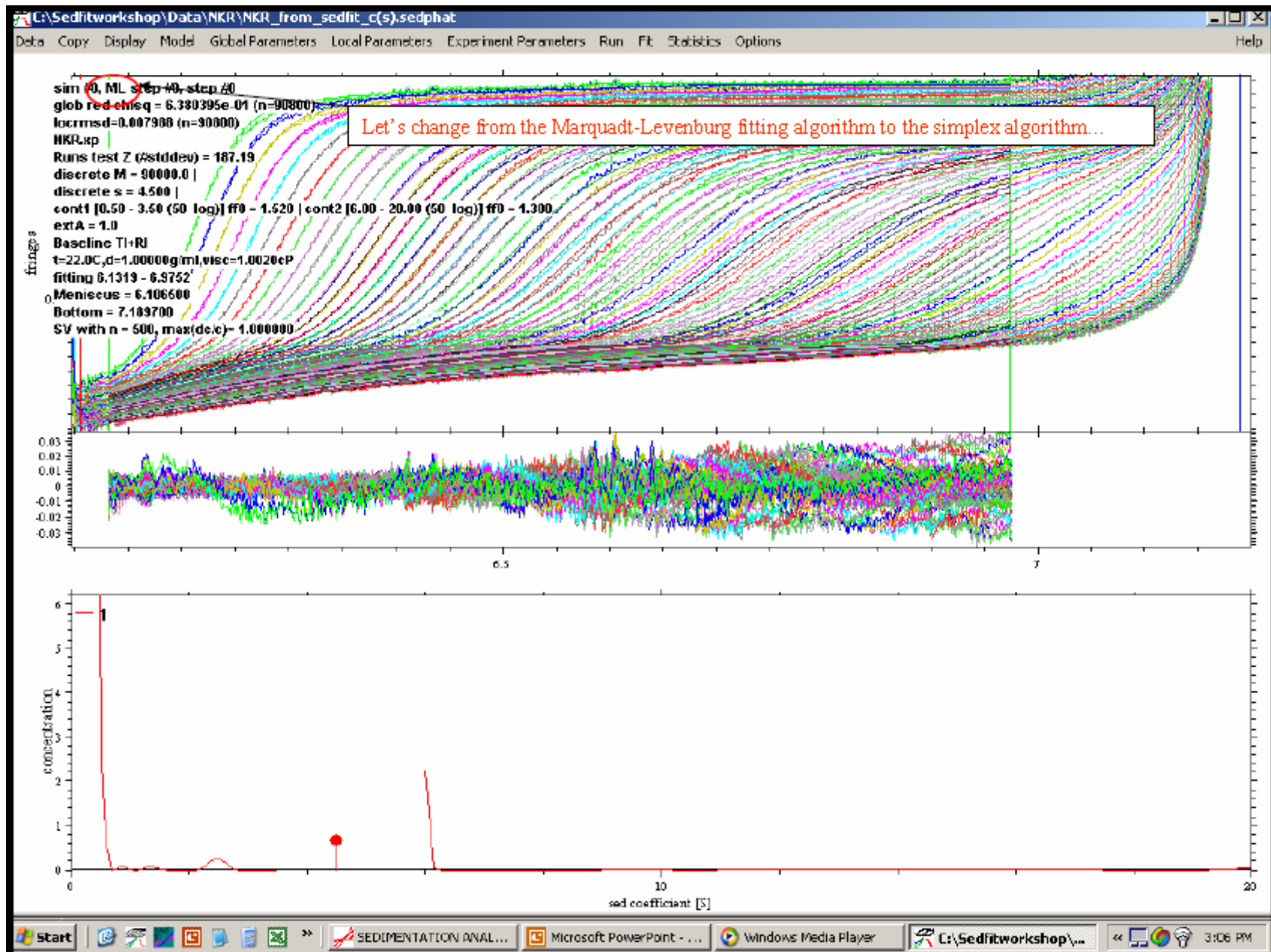


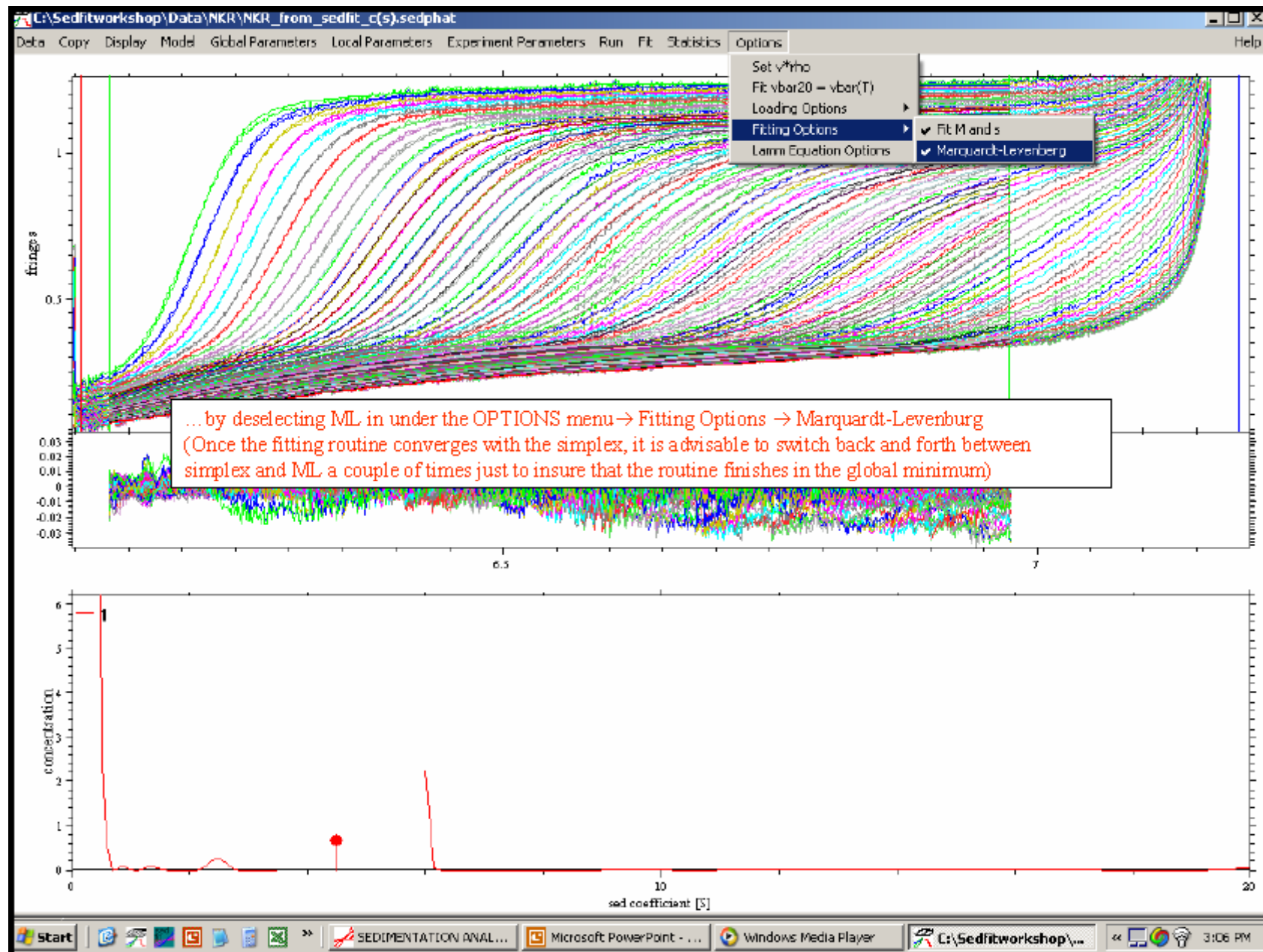


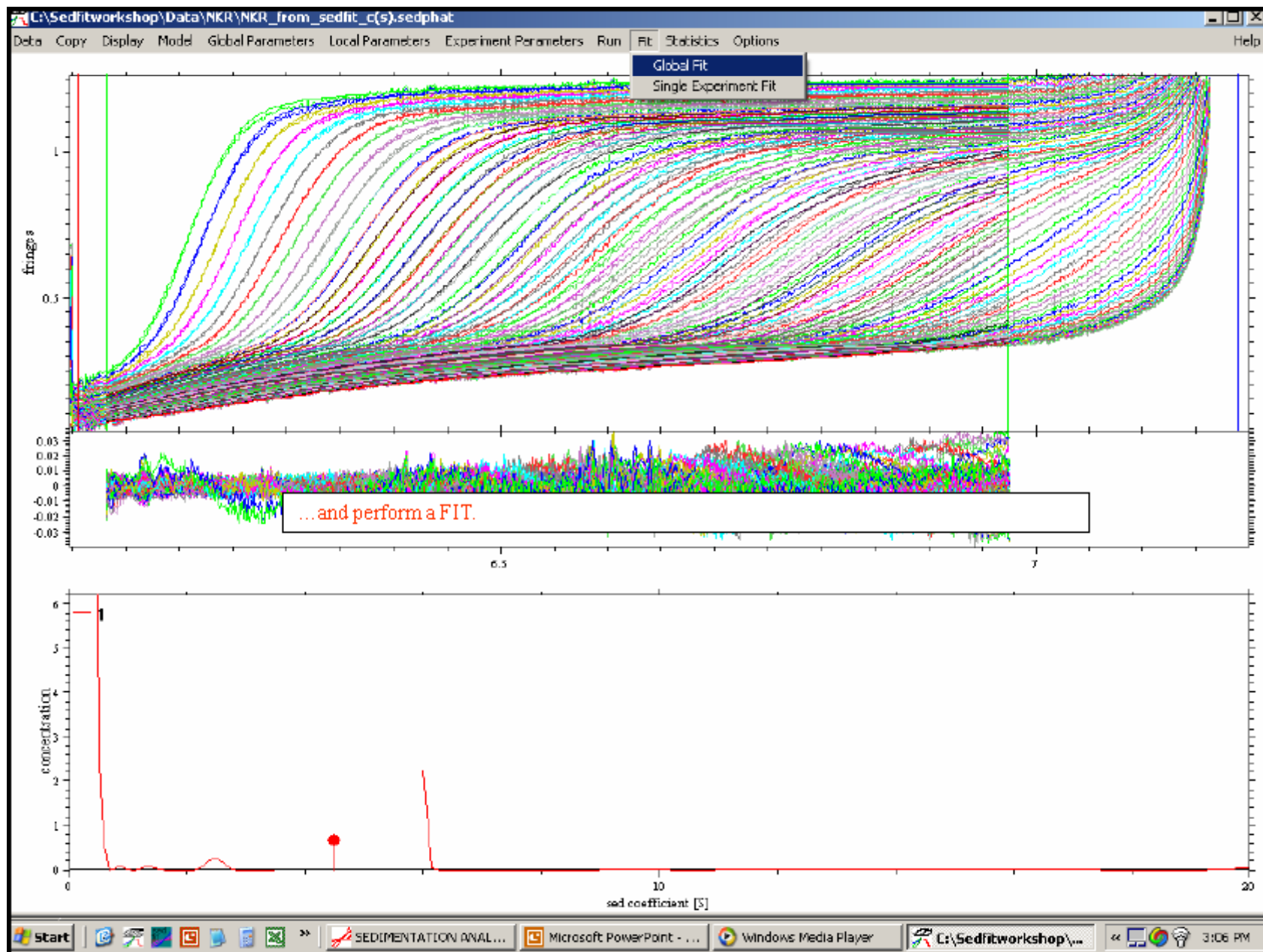


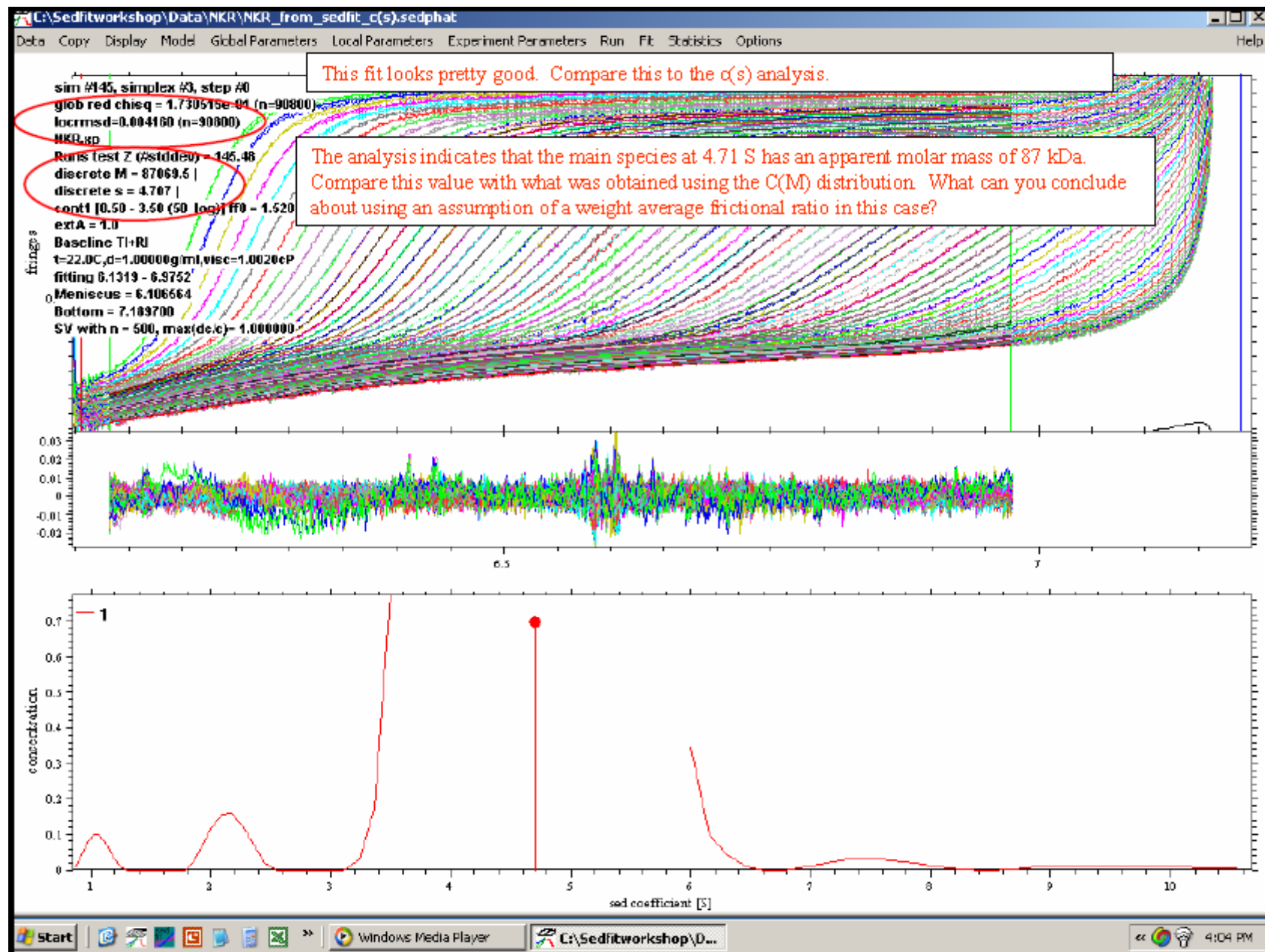


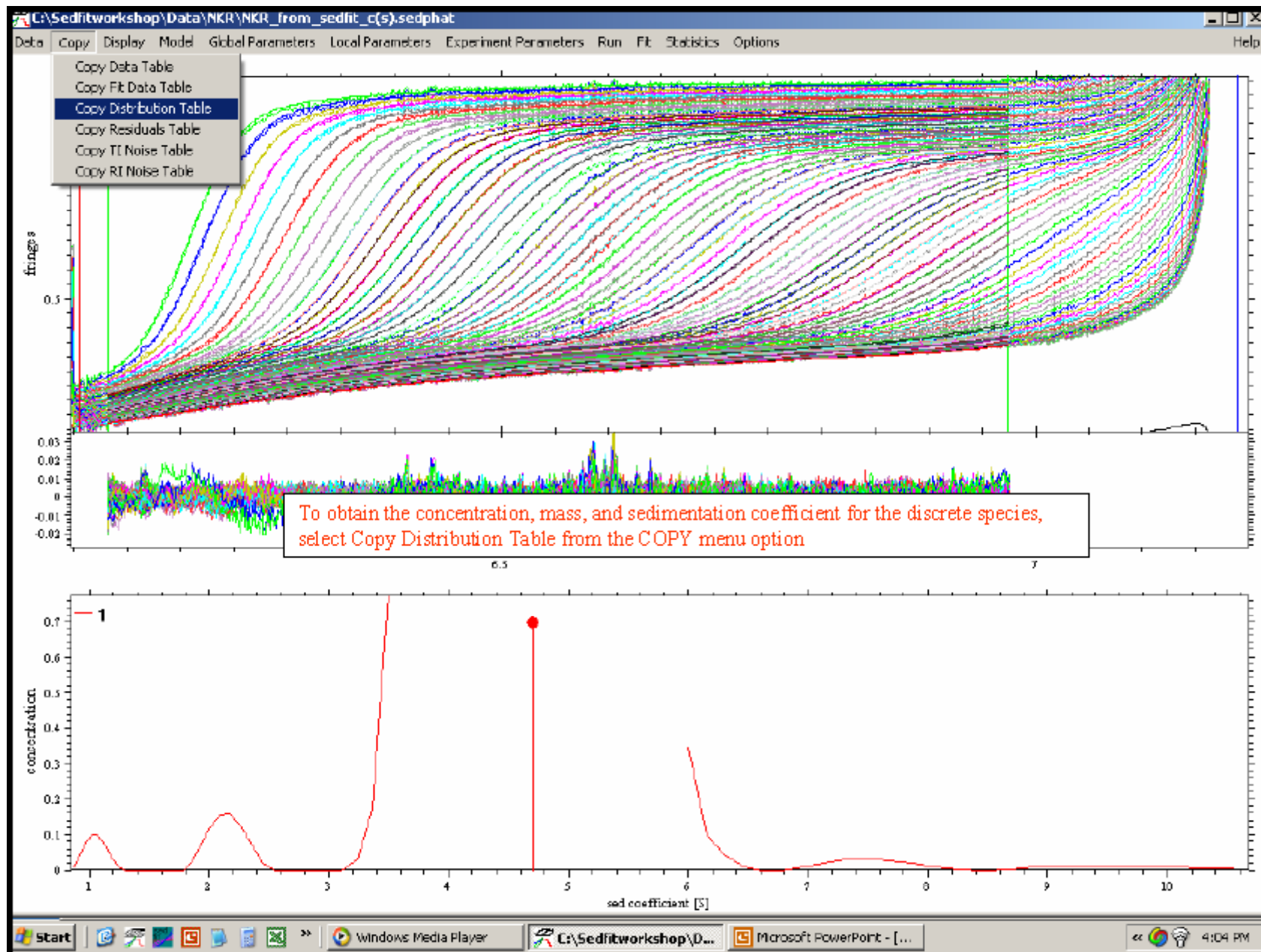


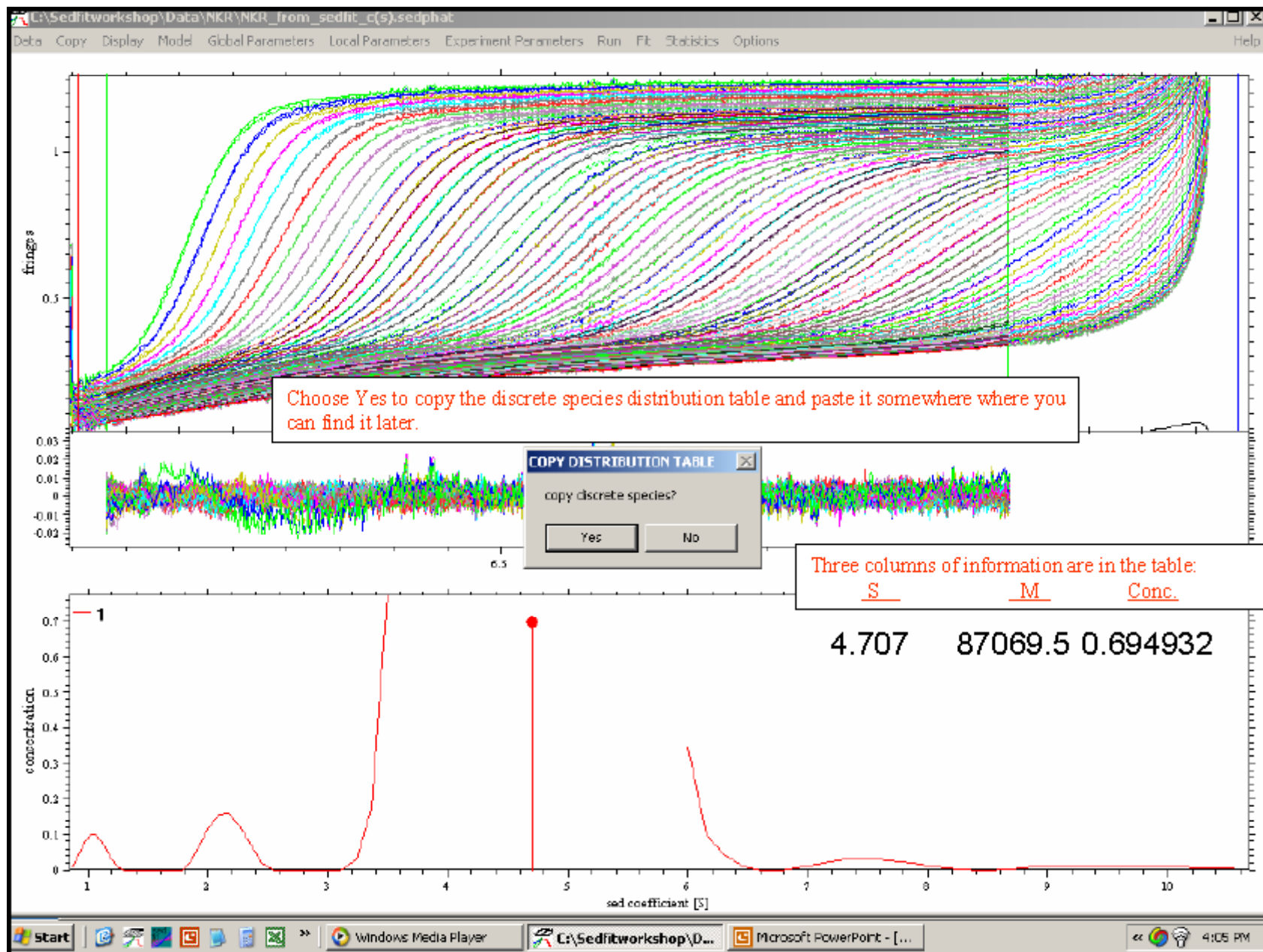








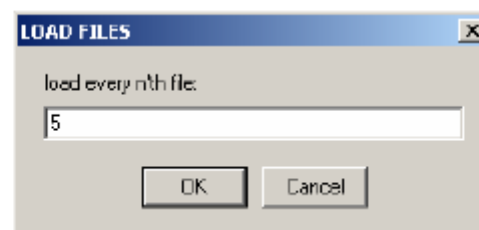
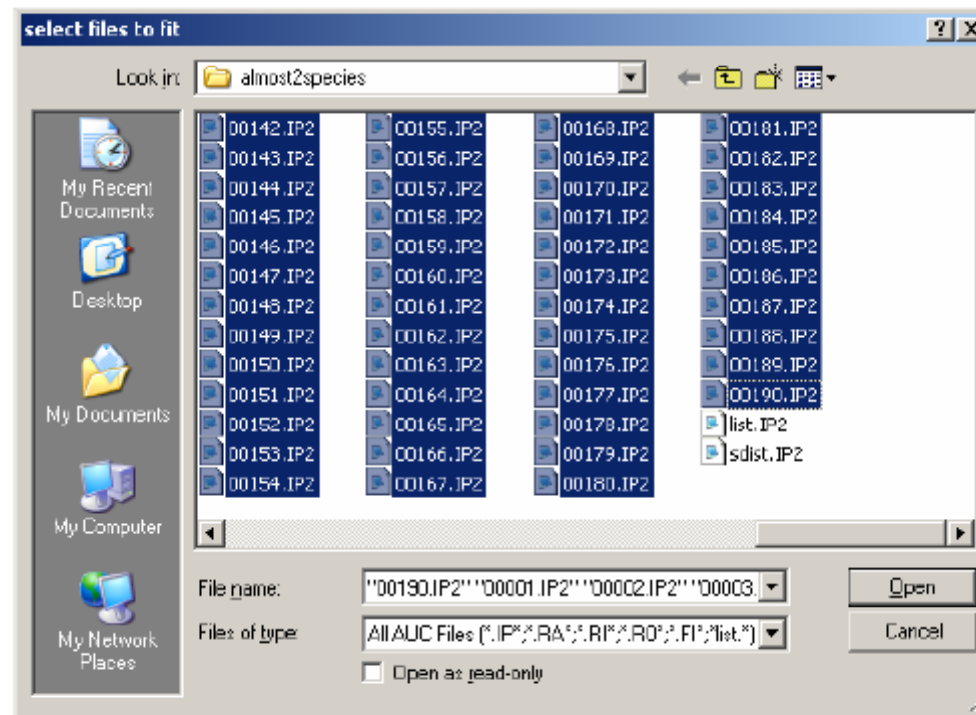


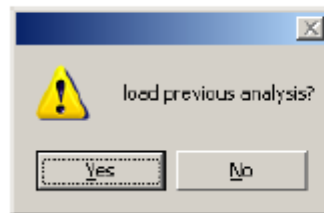


IV-1: Lamm equation modeling with a monomer-dimer system

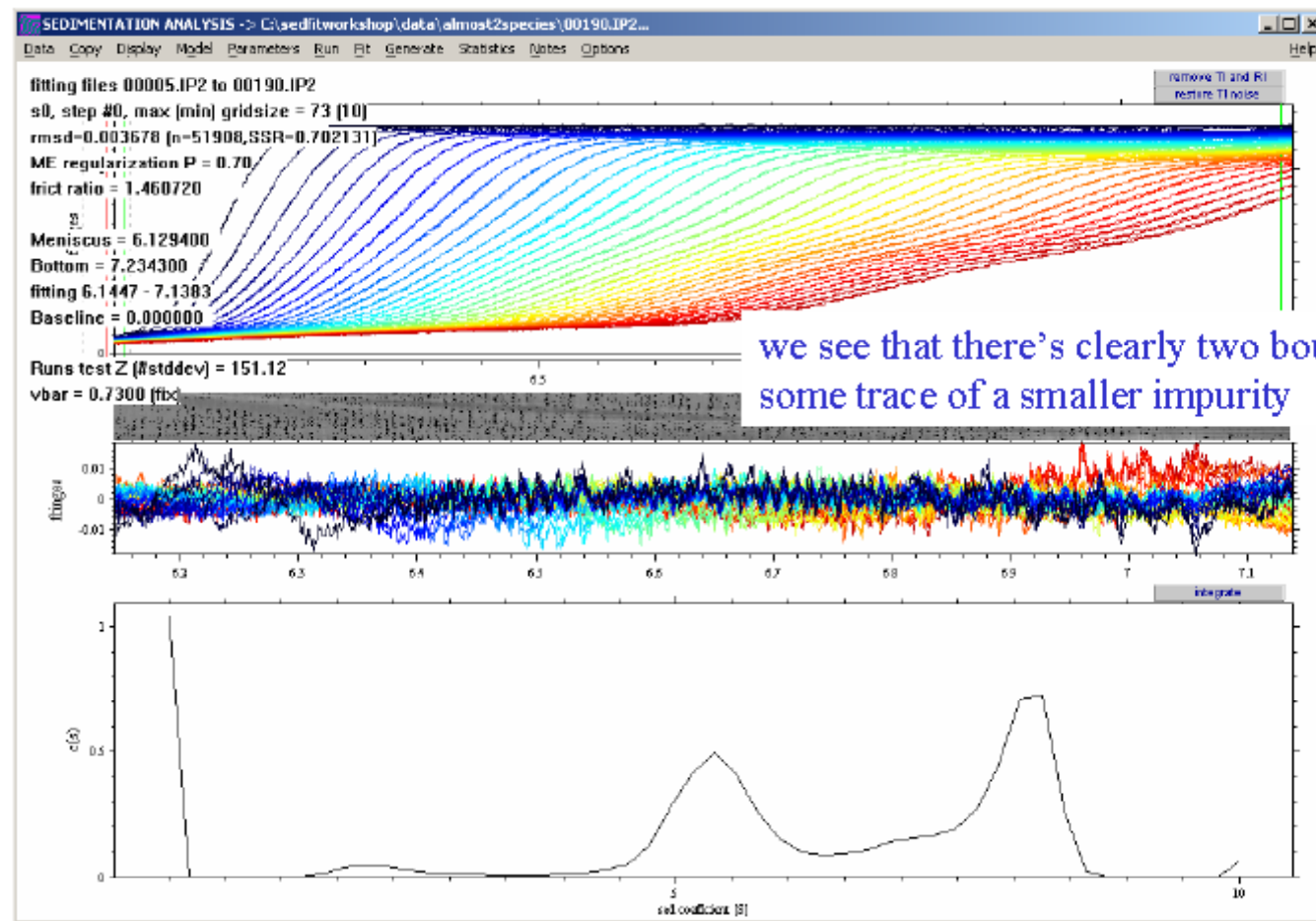
Let's suppose we have a protein ~ 100 kDa that, based on $c(s)$ seems to form dimers. There's also a smaller Mw impurity. The goal of this exercise is to make the transition from $c(s)$ in SEDFIT to set up an analysis of the same data as a interacting monomer-dimer system in SEDPHAT.

open SEDFIT, and load all files in the example “almost2species”

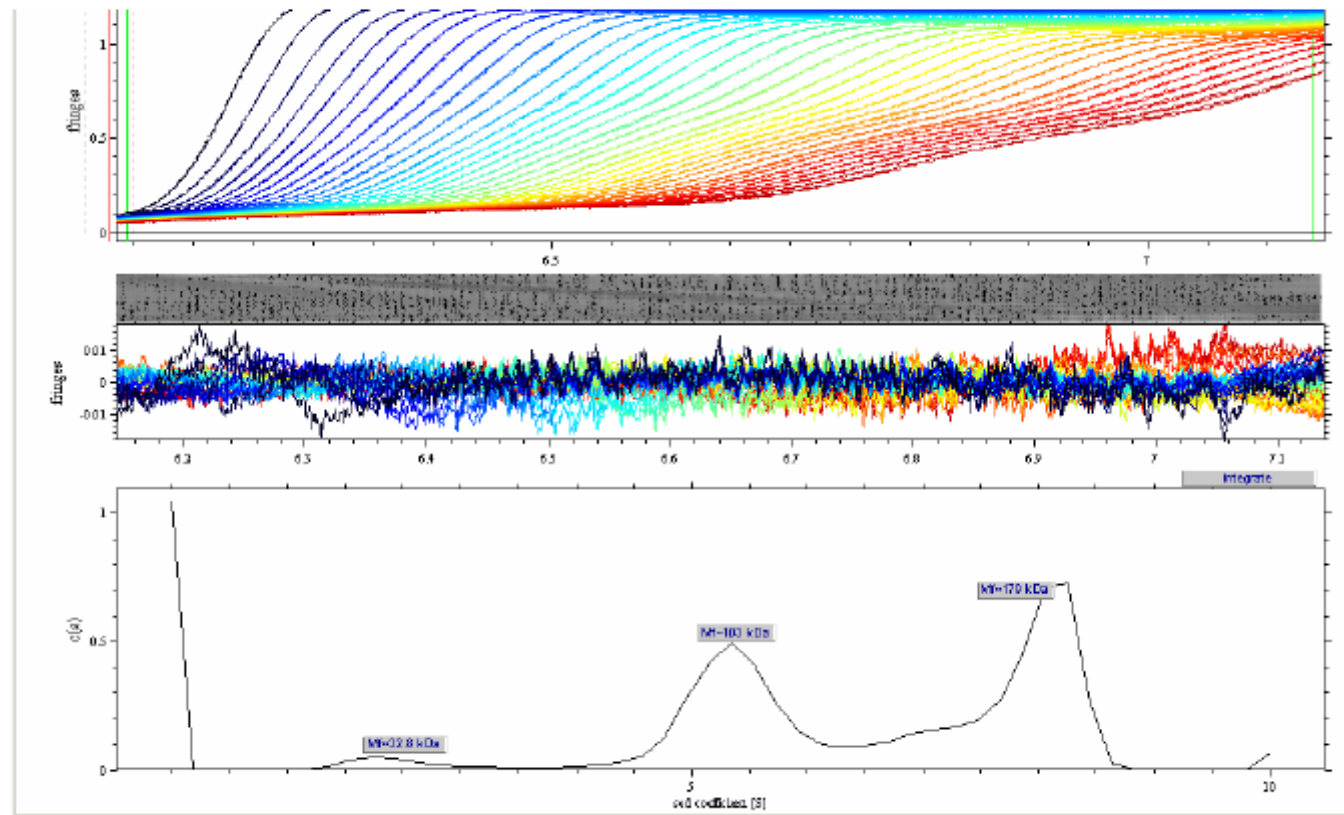




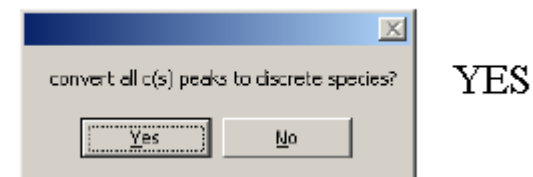
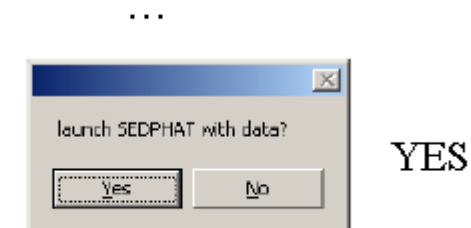
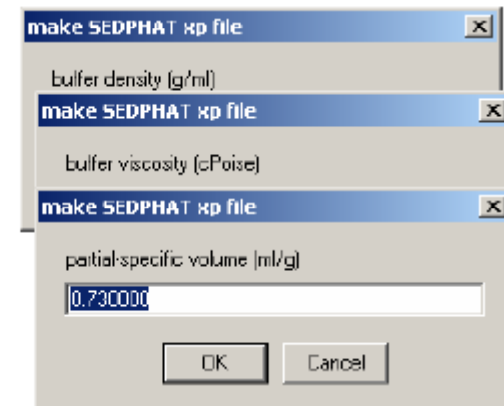
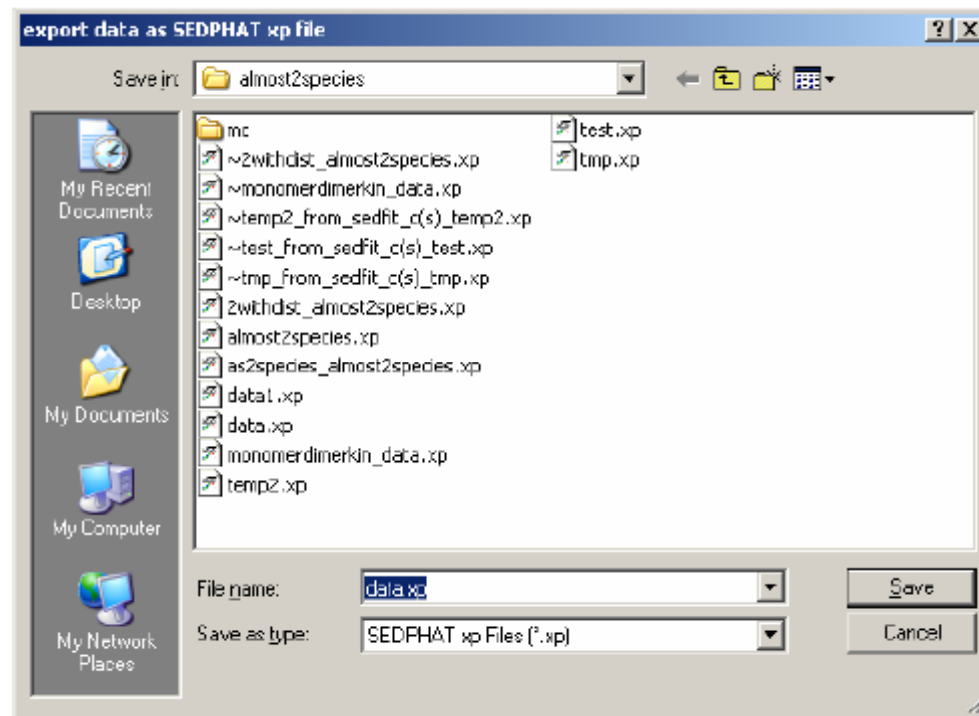
YES, then to a RUN command
 then do control-D [data range] and control-N [subtract TI+RI noise]
 This will reproduce a previously made standard $c(s)$
 [if you don't get to restore a previously saved fit, see tutorials on $c(s)$]



control-M shows the approximate Mw estimates: The 103 kDa should be the monomer, but the 179 kDa for what should be the dimer is too low a number, which means there might be excess boundary spread. Also, there's something in between the two peaks – this region should not be populated if we had really independent monomer and dimer



let's export this. Use the function Data → Export Data to SEDPHAT
Give the xp-file to be generated some filename, and default through all the
parameter boxes.



this will use the automatic integration in SEDFIT to generate starting guesses for SEDPHAT model treating all species as discrete species.

Parameters for Hybrid Local Continuous/Global Discrete Model

☒ sp.1 ☐ neg.	☒ species 2	☒ species 3	☐ species 4	☐ species 5
☒ M1 179053.92	☒ M2 102689.48	☒ M3 32834.937	☐ M4 0.00000	☐ M5 0.00000
☒ S1 7.842333	☒ S2 5.413440	☒ S3 2.531319	☐ S4 0.000000	☐ S5 0.000000
☐ species 6	☐ species 7	☐ species 8	☐ species 9	☐ species 10
☐ M6 0.00000	☐ M7 0.00000	☐ M8 0.00000	☐ M9 0.00000	☐ M10 0.00000
☐ S6 0.000000	☐ S7 0.000000	☐ S8 0.000000	☐ S9 0.000000	☐ S10 0.000000

☐ link molar mass in multiples of 1 for species from #2 to #4

☐ link τ values in multiples of 1 for species from #2 to #4

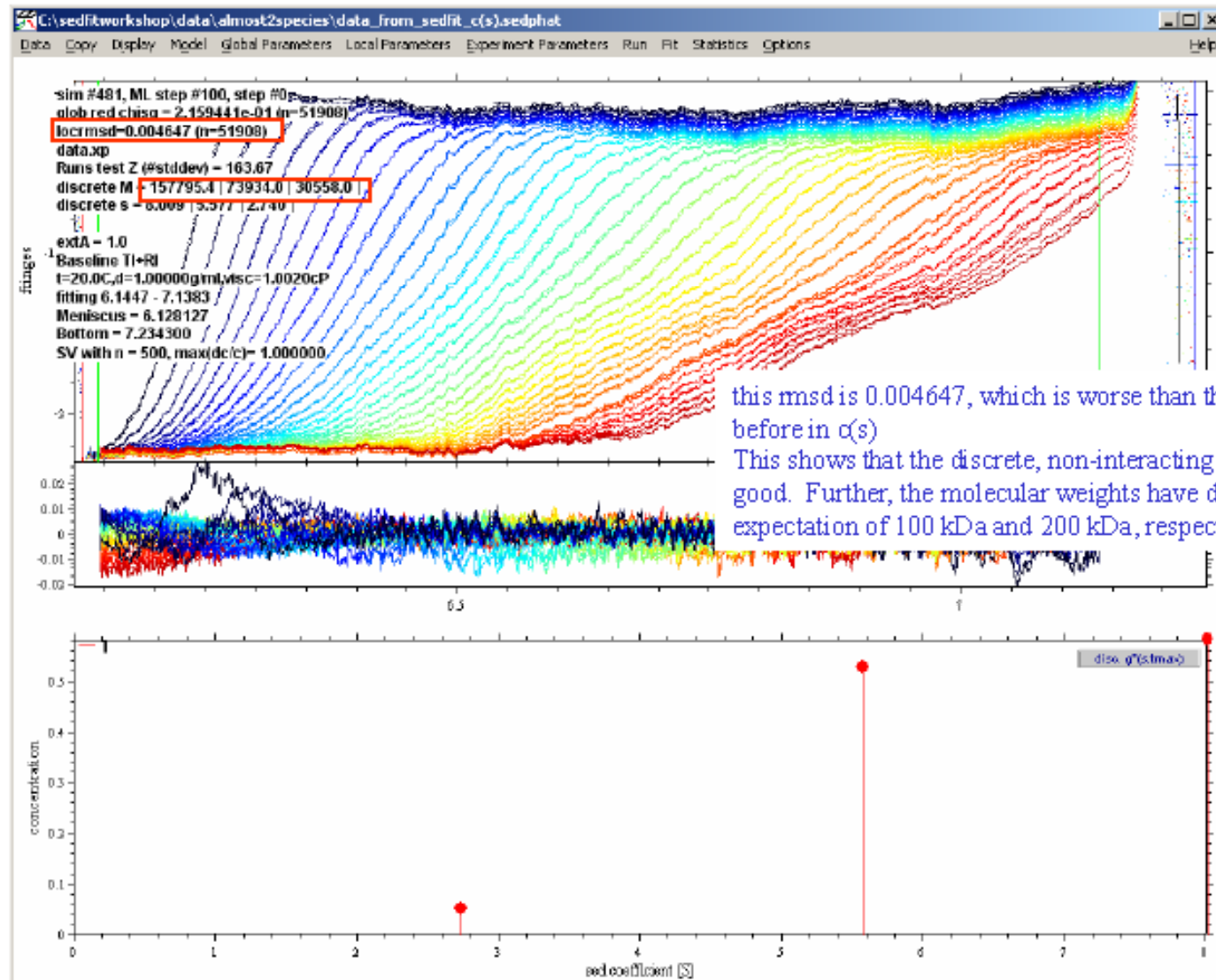
continuous	resolution	τ min	τ max	frictional ratio	fit f(R)	
☐ segment 1	1	☐ linear ☒ log	0.500	10.000	1.451	☐ fit f(R)
☐ segment 2	0	☐ linear ☒ log	0.500	10.000	1.300	☐ fit f(R)
☐ segment 3	0	☐ linear ☒ log	0.000	0.000	1.300	☐ fit f(R)

☒ with Tikhonov Regularization R 0.700 ☒ normalize distributions

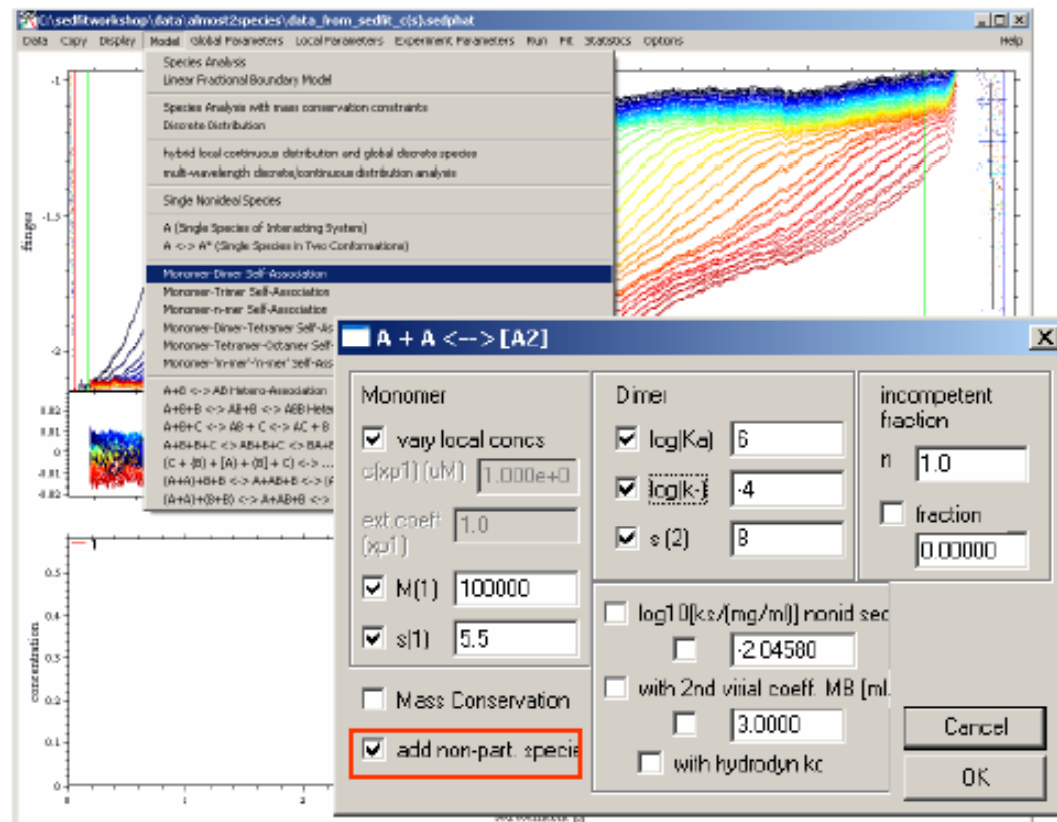
Cancel OK

accept, do RUN and then FIT

then fine-adjust the display by using control-D [set data range] and right-double click in the distribution window [full range distribution plot] and control-O [show last fit info]



Let's try the monomer-dimer model in order to account for finite reaction kinetics.



also check 'add non-part. species' box, since we will need to account for the slow sedimenting degradation product when fitting this data.

Then hit OK

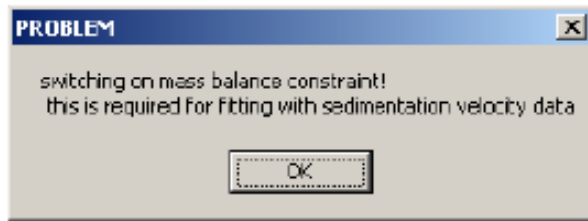
we use the following initial estimates:

$M1 = 100\text{kDa}$ [that should be the Mw from sequence, but we don't know the exact $\bar{v}$, therefore we'll float it here]

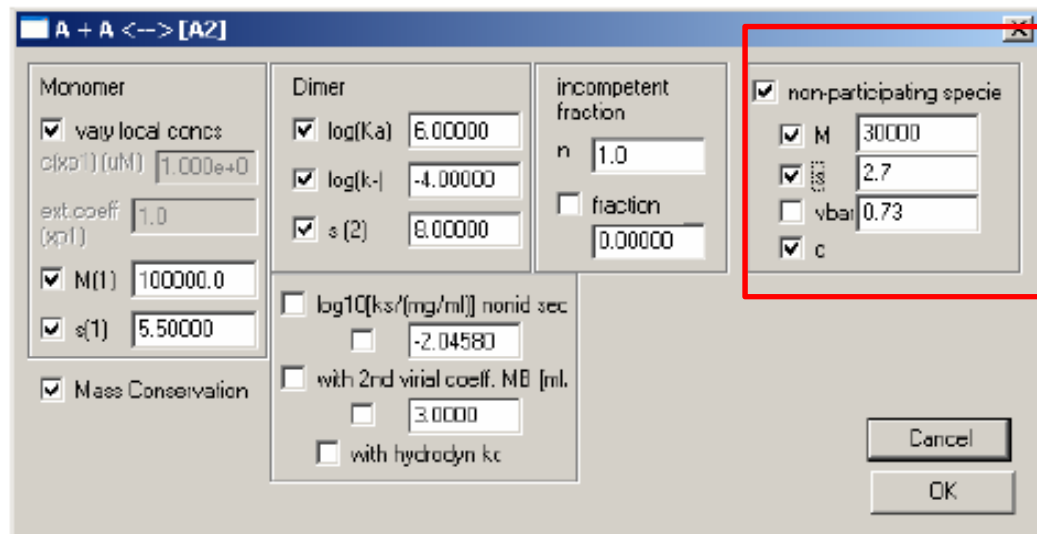
$s1 = 5.5, s2 = 8$ [these are the s-values from the discrete model, which should be excellent starting guesses and we'll float]

$\log(Ka) = 6$ [since we loaded $\sim \mu\text{M}$ concentration, and see roughly equal height of monomer and dimer peaks in $c(s)$, we should be close to K_d at these μM concs]

$\log(k-) = -4$ [the reaction must be fairly slow (i.e. -4 or slower) for us to resolve peaks in $c(s)$, but we want to put the starting guess at a value close to the range of -3 to -4 where the SV experiment is sensitive to the value of $\log(k_{\text{off}})$. Otherwise the optimization of this parameter would become difficult.]



this is no problem, all SV models obey mass balance.



because we had checked the 'add non-part. species' box, now an extended box shows up with a field where we enter the parameters of the non-participating species.

This will be 30kDa and 2.7 S, which are the estimates taken from the discrete species fit

Then hit OK



Now a input box pops up asking us how large the concentration of the non-participating species is. This is in signal units, and from the discrete species fit we can estimate a value of 0.05. (Or this can be obtained also from integrating the previous $c(s)$ around this contaminant peak).

One more set of parameters is required, due to the fact that in the interacting species models we want to deal with molar binding constants. Therefore, we need starting concentrations in molar units, as well as extinction coefficients to relate the molar concentration to signal.

Open the Experimental Parameters:

Experimental Parameters

(1) INTERFERENCE data for SEDVELOCITY

C:\sedfit\workshop\data\almost2species\data.xp (000051P2 ...)

Comment: cm/pixel: 0.00072706533, Sample

☒ active

noise: 0.0100

Centerpiece: 2

Pathlength: 1.200000

Rotor type: 4

☐ no backdiffusion needed

☒ fit baseline

☒ fit RI Noise

☒ fit TI Noise

☒ Meniscus: 6.1261

☐ Bottom: 7.2343

☐ redirect men./bot.: 1

☐ redirect st A: 1

☐ redirect st B: 1

☐ redirect st C: 1

For Associating Systems:

☒ extinction coefficient A: 275000

☐ extinction coefficient B: 0.0000

☐ extinction coefficient C: 0.0000

☐ partial boundary fitting

smin: 0.00

smex: 0.00

Cancel

OK

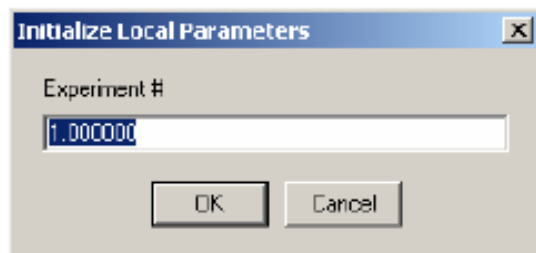
For the 'extinction coefficient A', enter a value of 275000. This is interference data, and based on the sensitivity for proteins of 3.3 fringes/mg/ml, we have the 'interference extinction coefficient' $2.75 \times M_w$.

Note we also have a pathlength of 1.2 cm, which ~~will allow SEDPHAT~~ to correct for the 12 mm thickness of the centerpiece used.

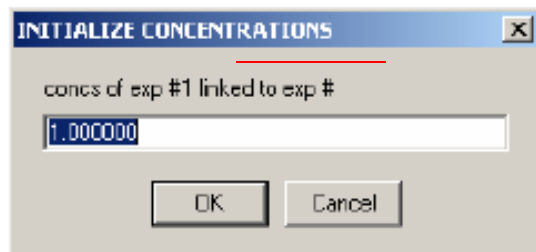
If this was absorbance data, we would enter here the conventional molar extinction coefficient (OD/M/cm), just as taken, for example, from SEDNTERP.

Click on OK

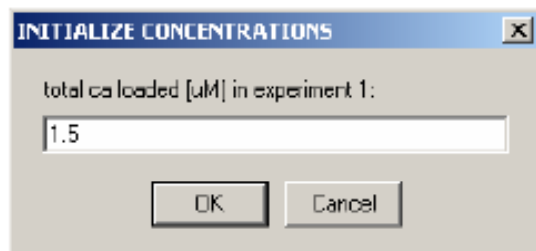
Next, choose the “concentrations (indep + associating species)” from the menu.



we're referring to xp #1 (have only 1 loaded anyway...)

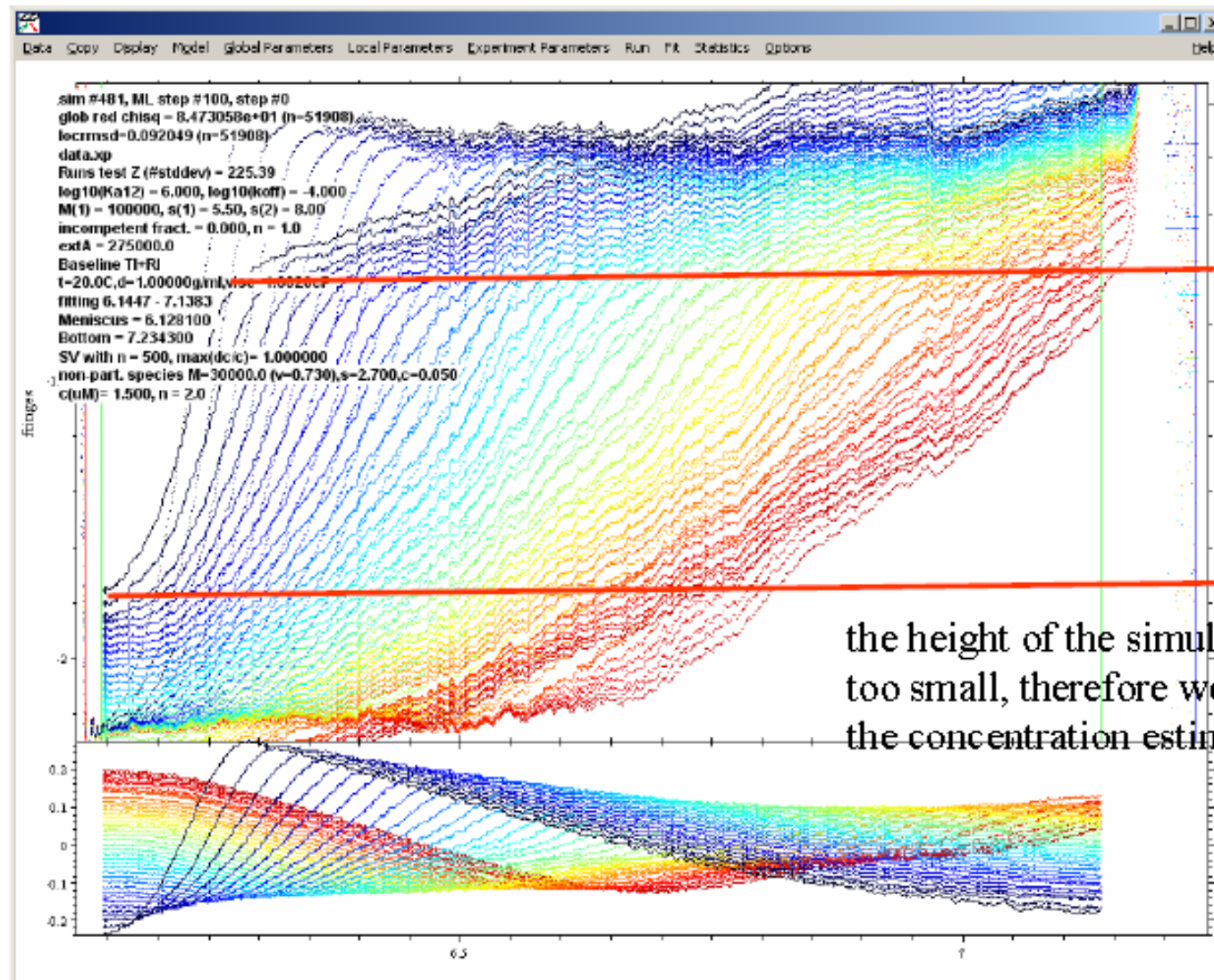


we'll not redirect this xp's concentration to another xp (again, since we only have one loaded – this function would make sense, e.g., if we had absorbance data from the same cell loaded as xp2 such that we should constrain the concentrations to be the same – ‘redirected’. See the sedimentation equilibrium tutorials, where heavy use of this option is made.)

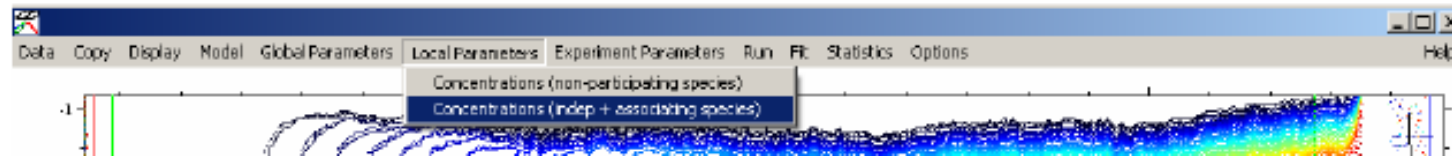


I believe we loaded a concentration of ~1.5 uM in this cell, therefore we enter the value 1.5.

Do a RUN to test the starting guesses.



Again, choose the “concentrations (indep + associating species)” from the menu.

A screenshot of a dialog box titled "Initialize Local Parameters". It has a field for "Experiment #" with the value "1.000000". There are "OK" and "Cancel" buttons at the bottom.

we're referring to xp #1 (have only 1 loaded anyway...)

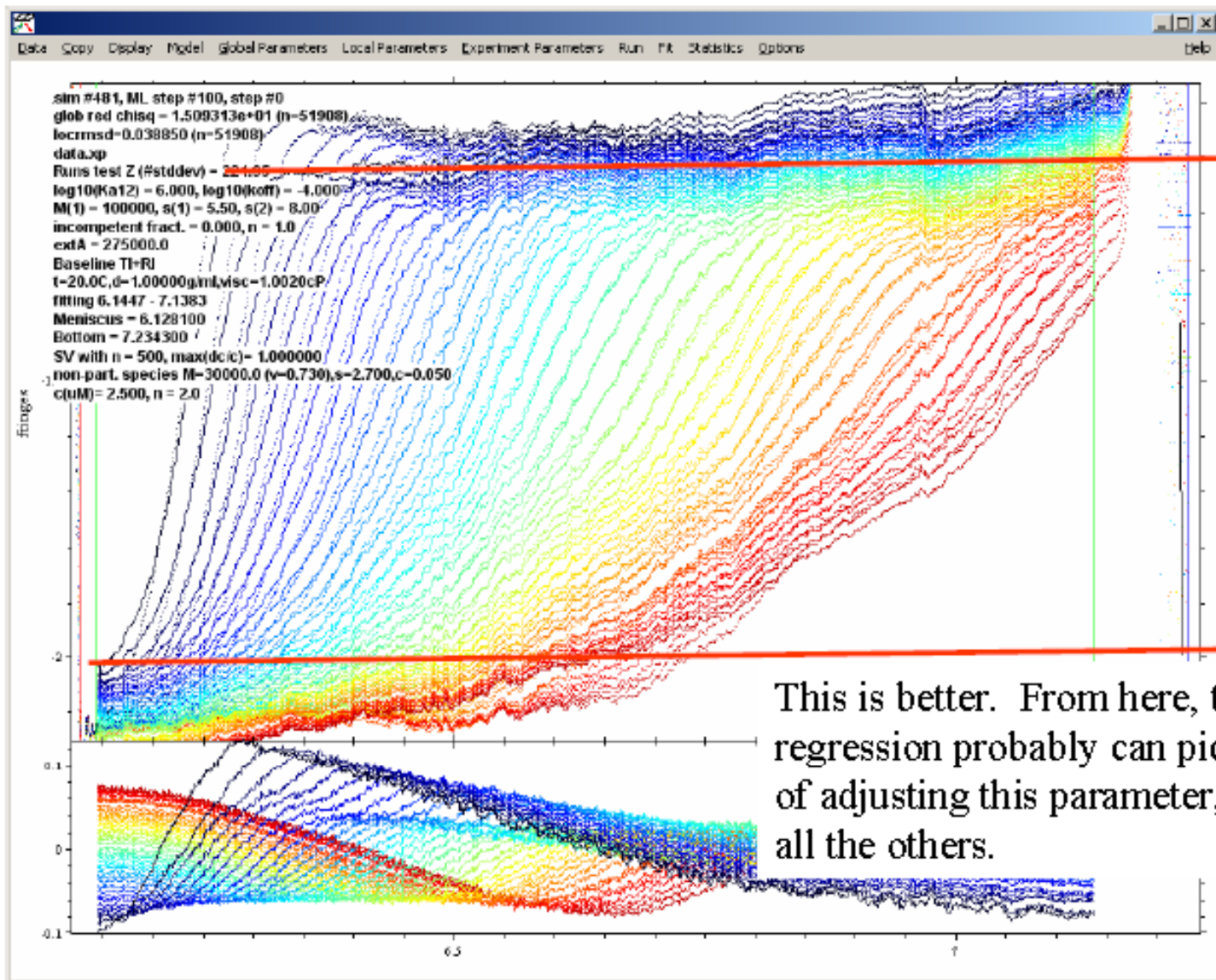
A screenshot of a dialog box titled "INITIALIZE CONCENTRATIONS". It has a field for "concs of exp #1 linked to exp #" with the value "1.000000". There are "OK" and "Cancel" buttons at the bottom.

we'll not redirect this xp's concentration to another xp

A screenshot of a dialog box titled "INITIALIZE CONCENTRATIONS". It has a field for "total cs loaded [uM] in experiment 1:" with the value "2.5". There are "OK" and "Cancel" buttons at the bottom.

Let's try a value of 2.5 uM.

Do a RUN to test the starting guesses.



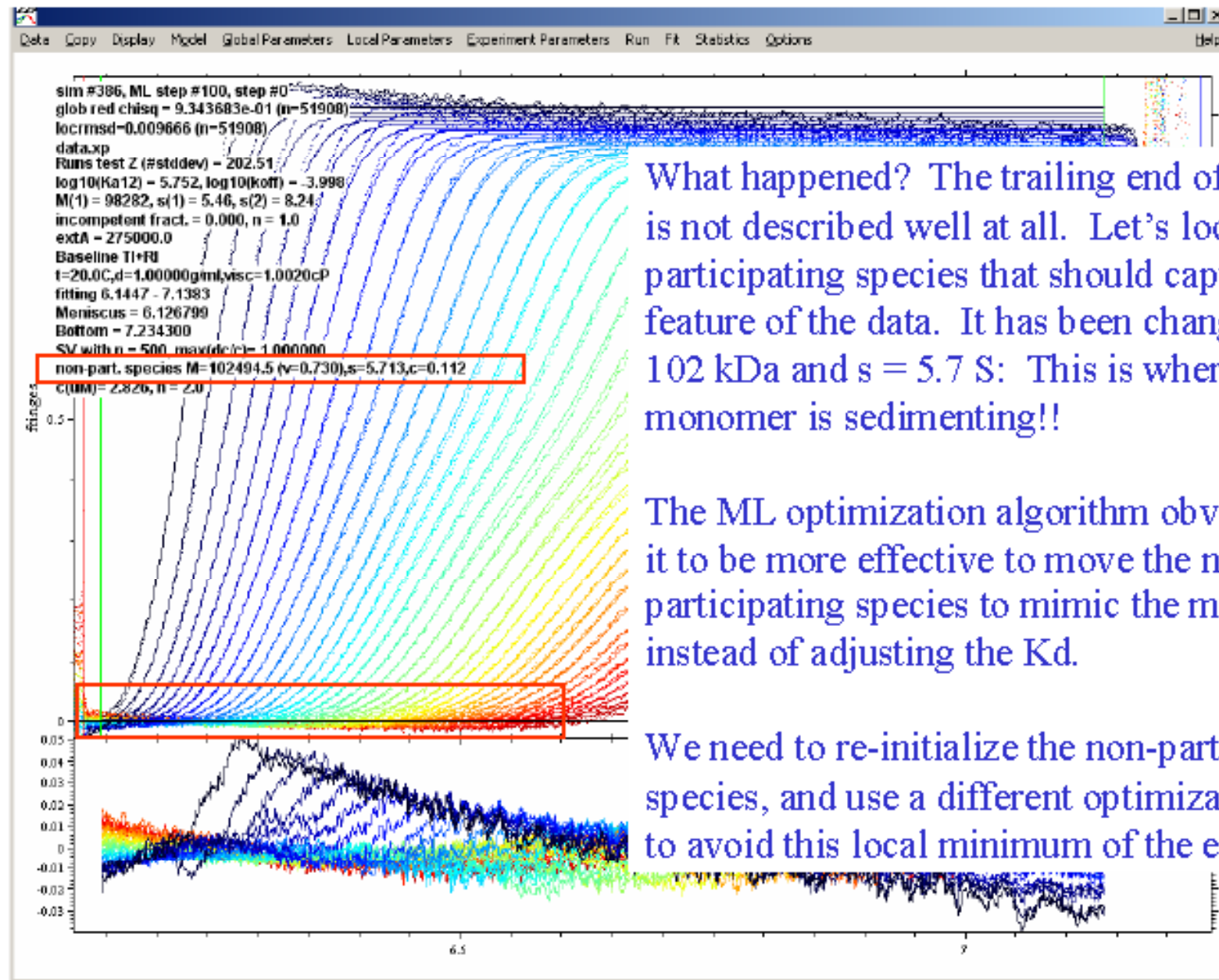
Now click on FIT. Note, the 'ML' in the first line indicates we are currently using Marquardt-Levenberg optimization, which will need to do a large number of simulations to initialize the gradient information and seems not to do much before updating the screen... nevertheless, this is a very efficient algorithm overall.

If your copy of SEDPHAT is configured to use the Simplex or the Simulated Annealing algorithm for minimization, the following sequence of events ~~will be different. There are many different equally~~ valid options how to do the fitting.

As it turns out, the following documentation will not be of a straightforward fit, but one where we have to avoid getting trapped into local minima. We use a combination of fitting algorithms to get to the best fit in a shallow minimum. (Remember the error surfaces of the banana function and the mountain range.) The details of the following are probably not reproducible, because the Simplex routine involves seed points generated randomly, and therefore it will slightly differ each time. However, it should be straightforward to apply the same strategy to your particular situation, as well as to other fits.

From here on, this is more an example on switching between fitting algorithms.

After the fit has stopped, hit control-N [subtract noise] and control-O [show best-fit info again]. This is where I ended up with after the fit. This does not look very good, yet.



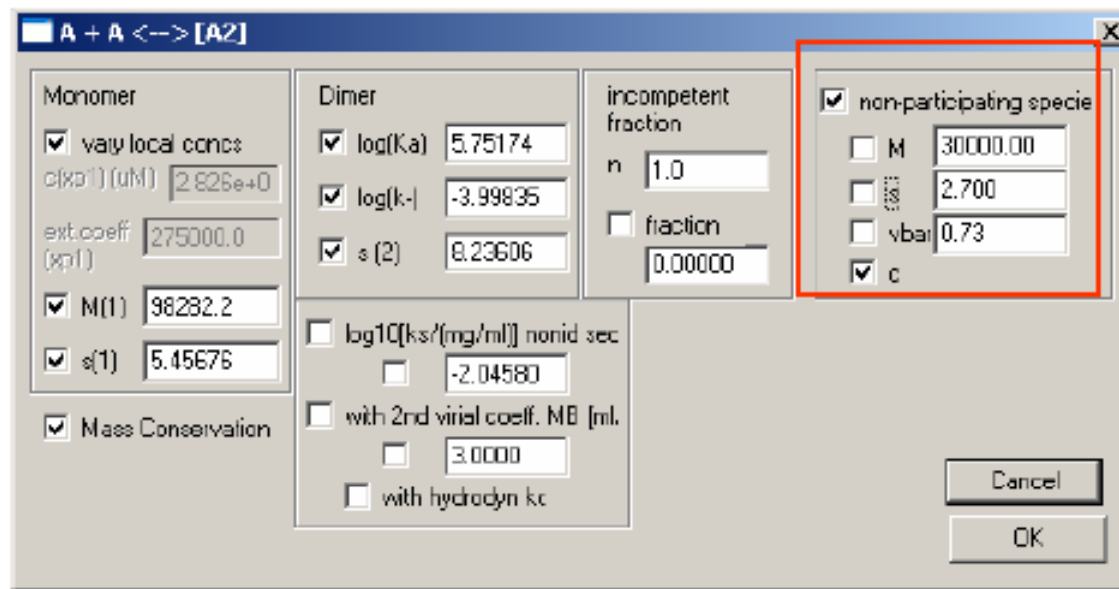
What happened? The trailing end of the boundary is not described well at all. Let's look at the non-participating species that should capture this feature of the data. It has been changed to $M = 102$ kDa and $s = 5.7$ S: This is where the monomer is sedimenting!!

The ML optimization algorithm obviously found it to be more effective to move the non-participating species to mimic the monomer, instead of adjusting the K_d .

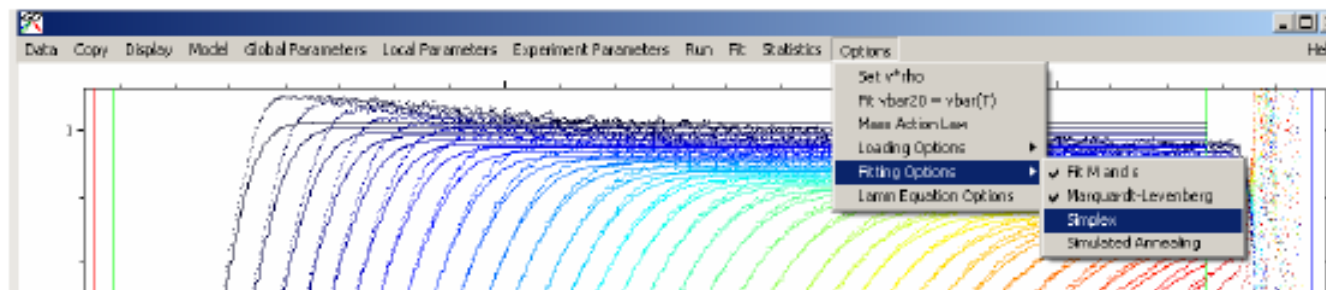
We need to re-initialize the non-participating species, and use a different optimization strategy to avoid this local minimum of the error surface.

in the Global parameters box, enter $M = 30000$ and $s = 2.7$

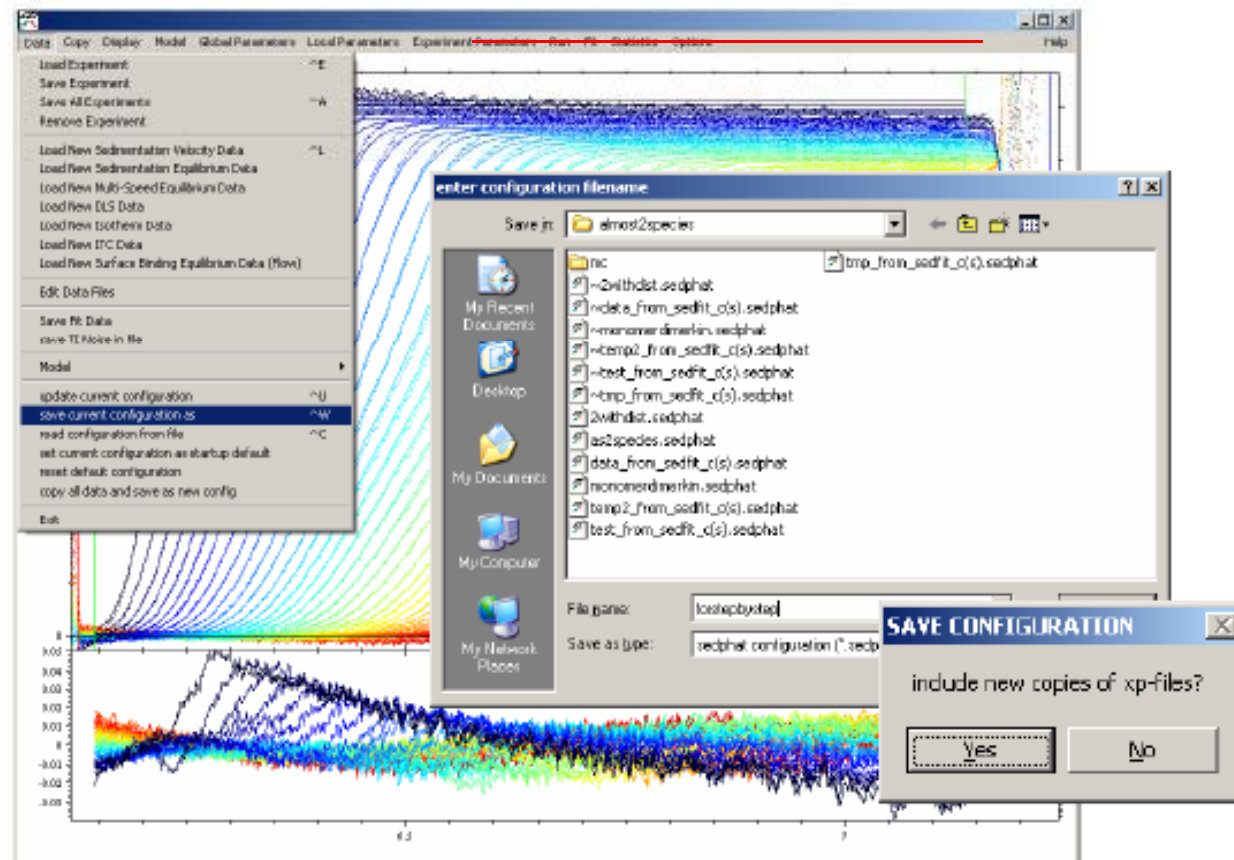
Also, we want to keep these parameters fixed now (we'll set them free later).



click on the “Simplex” menu function. This will toggle off Marquardt-Levenberg and
switch on the Simplex algorithm

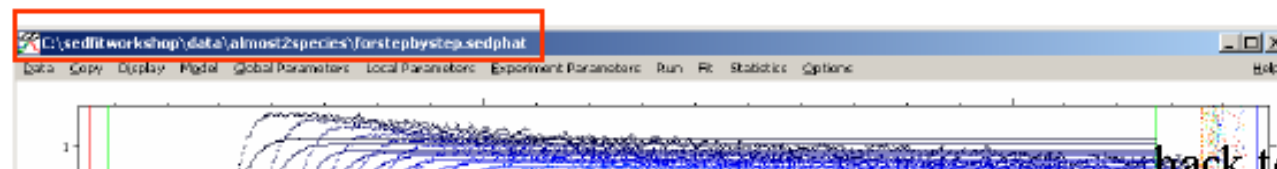


Also, just in case we need to go back, let us save this state of the analysis.
In the Data menu, click on “save current configuration as”, and enter a filename.



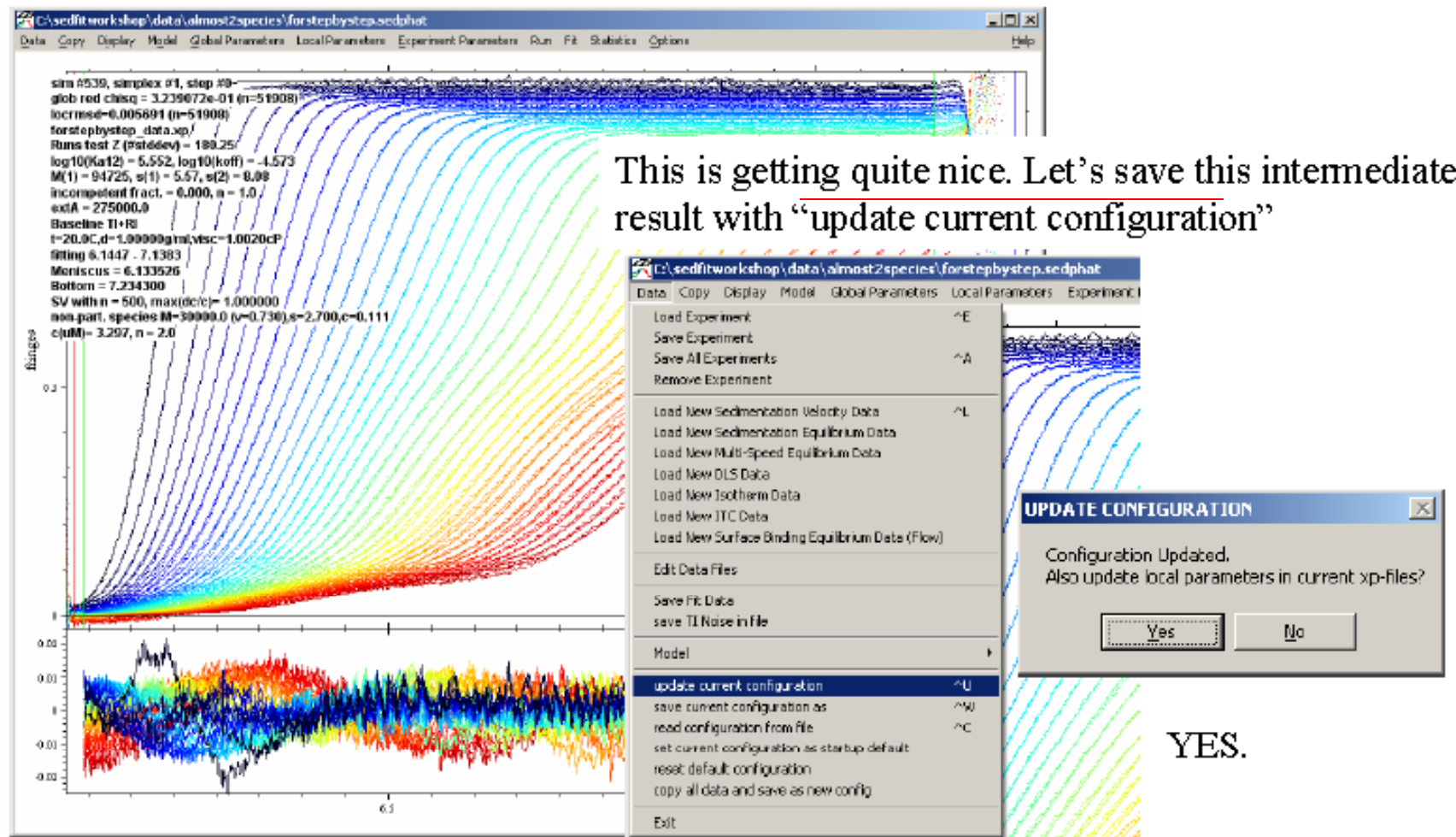
Yes, let's also
save the xp file.

This way, we can recreate this analysis later, and we'll be able to tell from the title bar the filename.

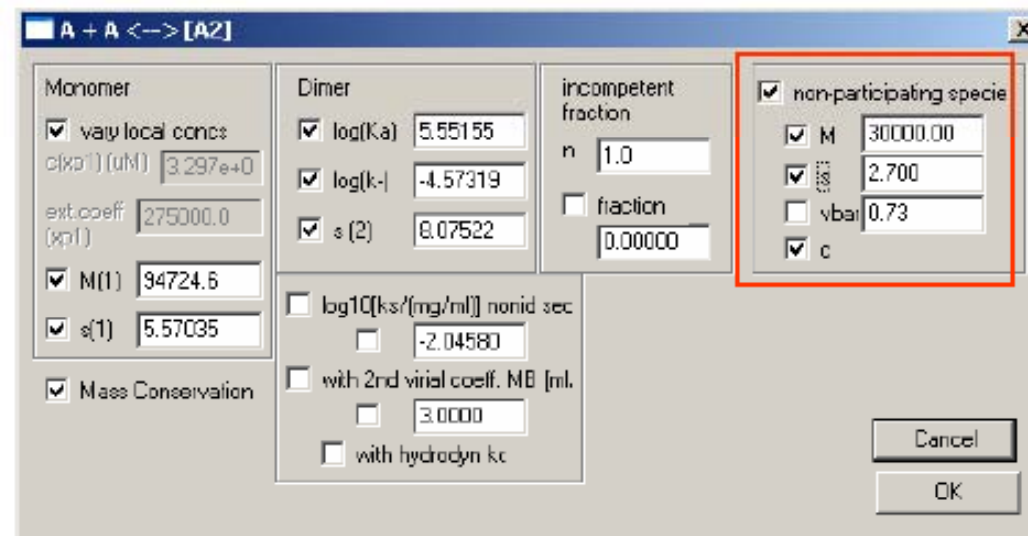


back to the fitting problem...

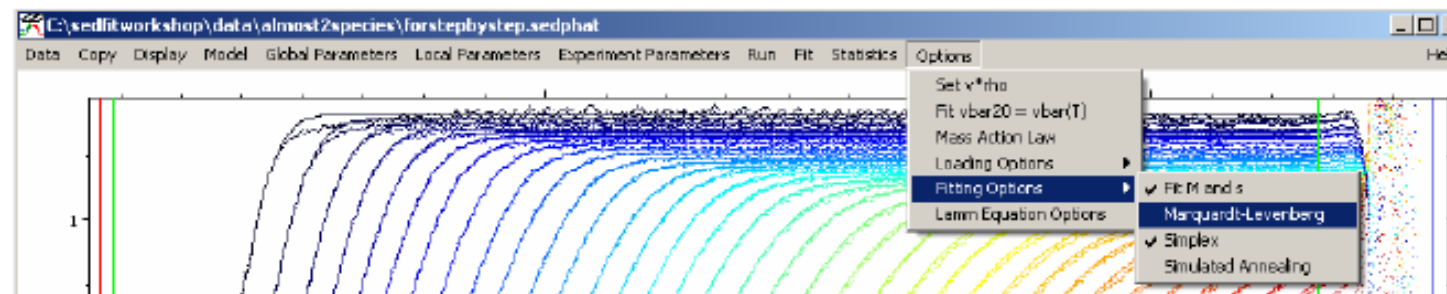
We could just let this go until it stops. I recommend that. However, being impatient, after the rmsd of the fit has dropped below 0.006, I interrupted this by hitting the space bar *once*. After control-N and control-O to get a good display, I find:



go back to the Global parameters box, and **float M and s of the non-participating species**

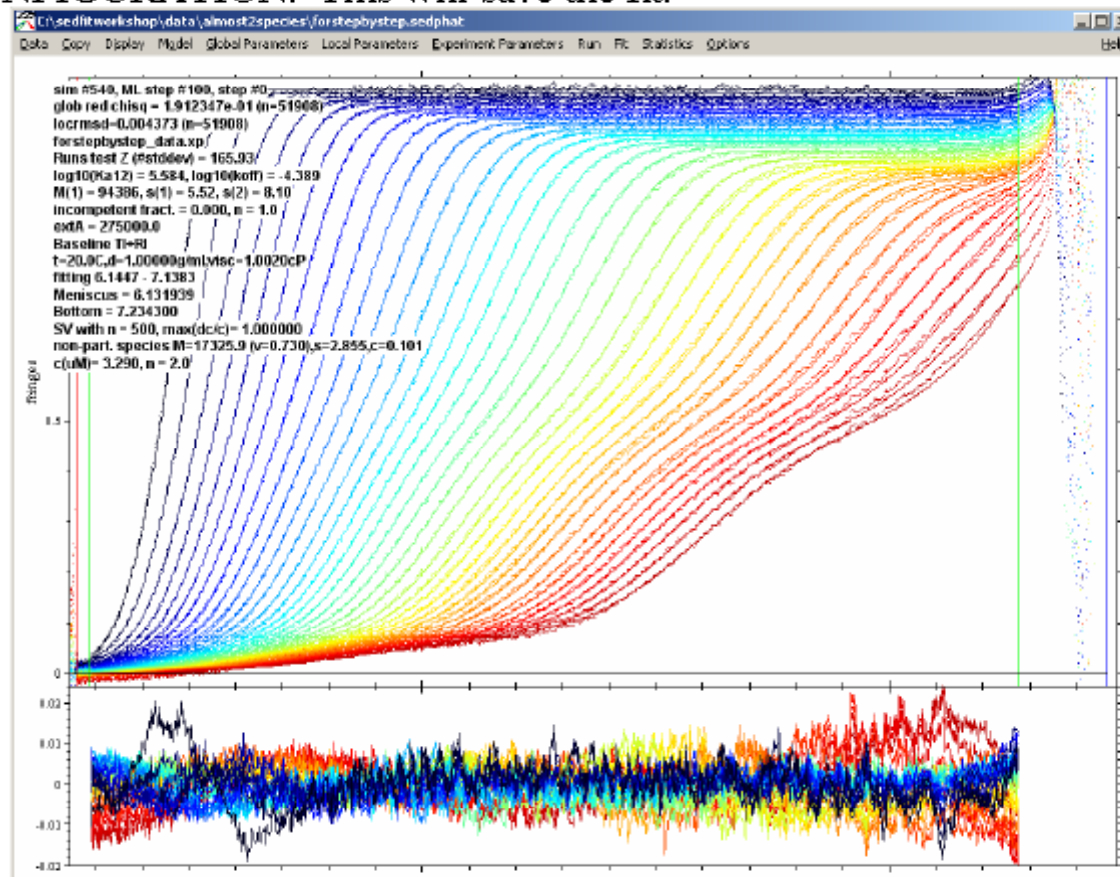


Also, we change back to Marquardt-Levenberg, to take advantage of its better ‘homing-in’ capability. Click on the “Marquardt-Levenberg” menu function to toggle.



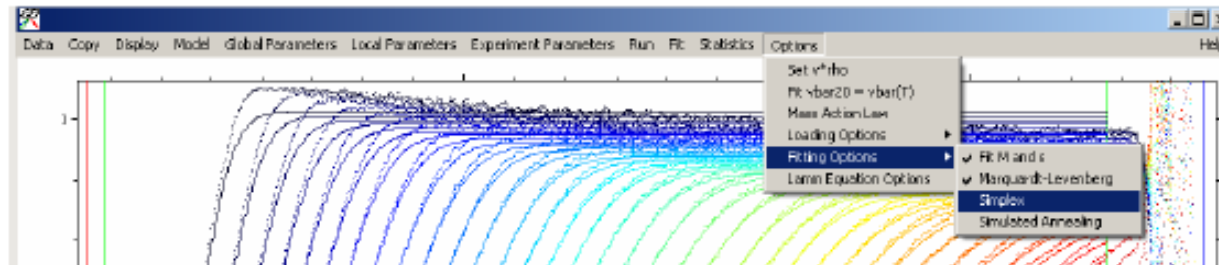
Then do a FIT

after a minute or two, we find the following best-fit (I used control-N and control-O to optimize display). use control-U or the menu function to UPDATE CURRENT CONFIGURATION. This will save the fit.



this fit does not look optimal, yet, as judged by the systematic residuals and misfit in the faster boundary. Therefore, we switch back to Simplex optimization and try to improve further.

click on the “Simplex” menu function. This will toggle off Marquardt-Levenberg and switch on the Simplex algorithm

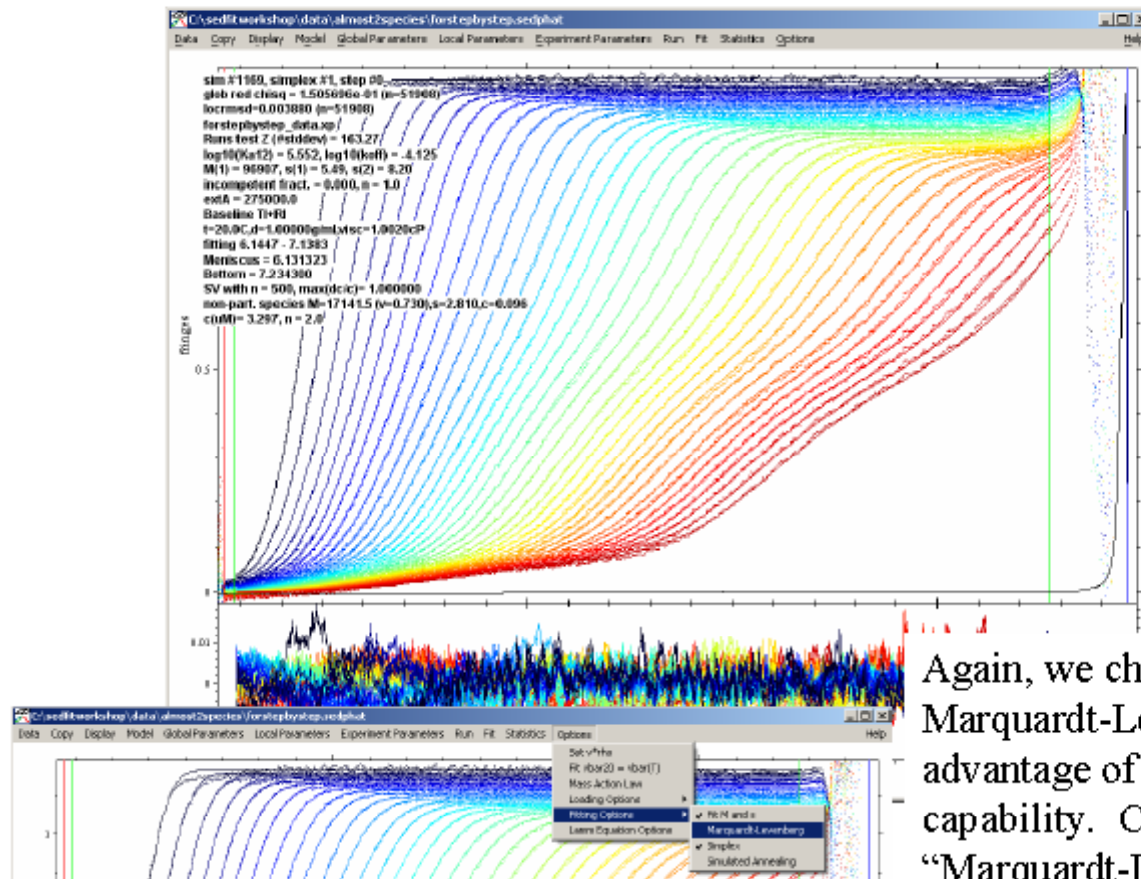


Then click on **FIT**

I saw initially a big improvement, indicating that the Marquardt-Levenberg again had been trapped in a small local minimum, from which the Simplex helped us escape. But after a while, convergence seemed to slow down very much, such that I again hit the space bar *once* to interrupt the fit.

[Hitting it once just stops the fit. It will then do a last simulation using the so-far best fitting parameters. If you inadvertently hit it twice, you are aborting this last simulation, and the fitted lines will be far off. However, you can rescue this by using a RUN command to re-establish the best-fit curves.]

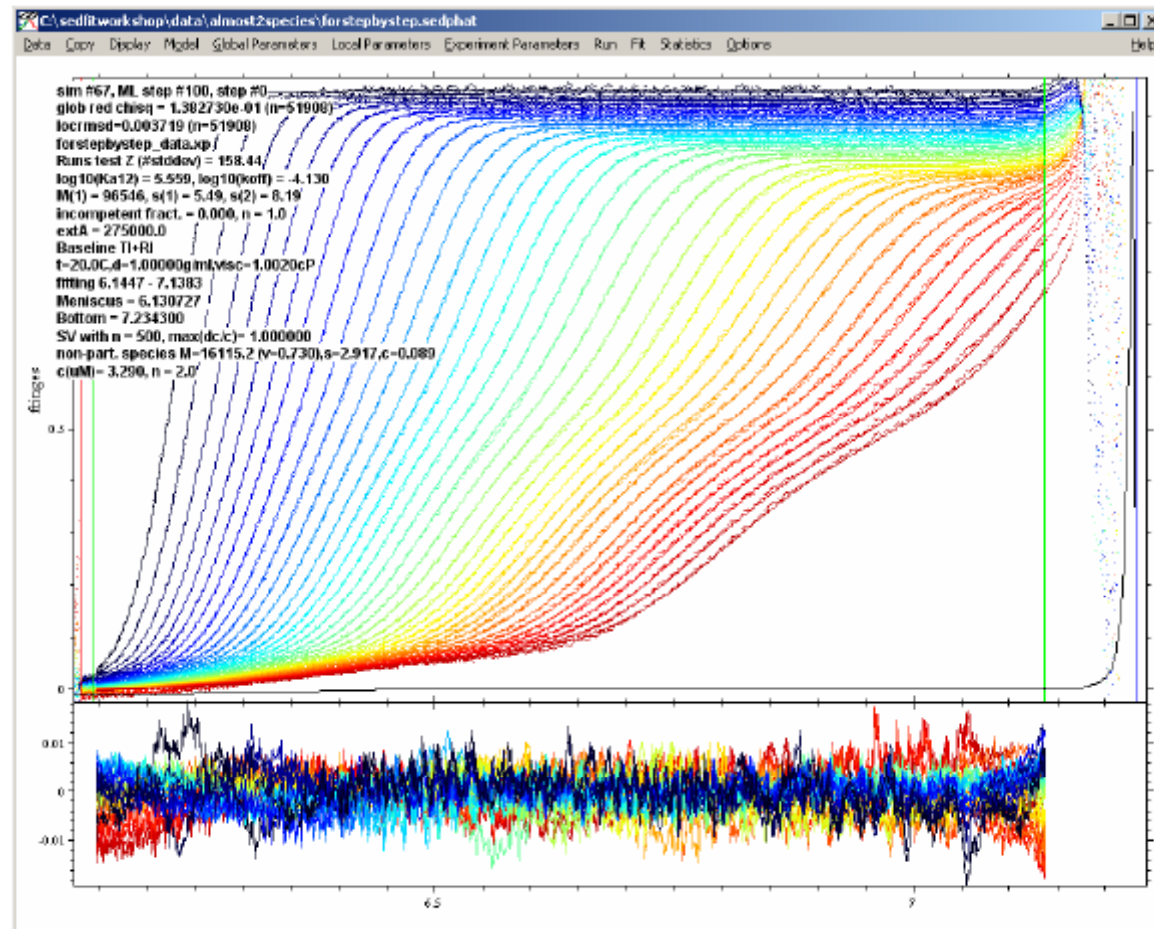
This is what I found. Use control-U or the menu function to UPDATE CURRENT CONFIGURATION. This will save the fit. Always also save the xp-file with it.



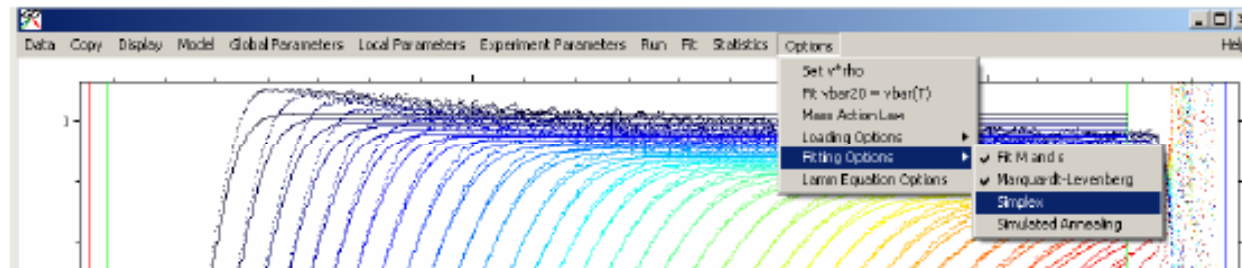
Again, we change back to Marquardt-Levenberg, to take advantage of its better ‘homing-in’ capability. Click on the “Marquardt-Levenberg” menu function to toggle.

Then do a FIT

This improved it slightly (judged by rmsd). Use control-U or the menu function to UPDATE CURRENT CONFIGURATION. This will save the fit.



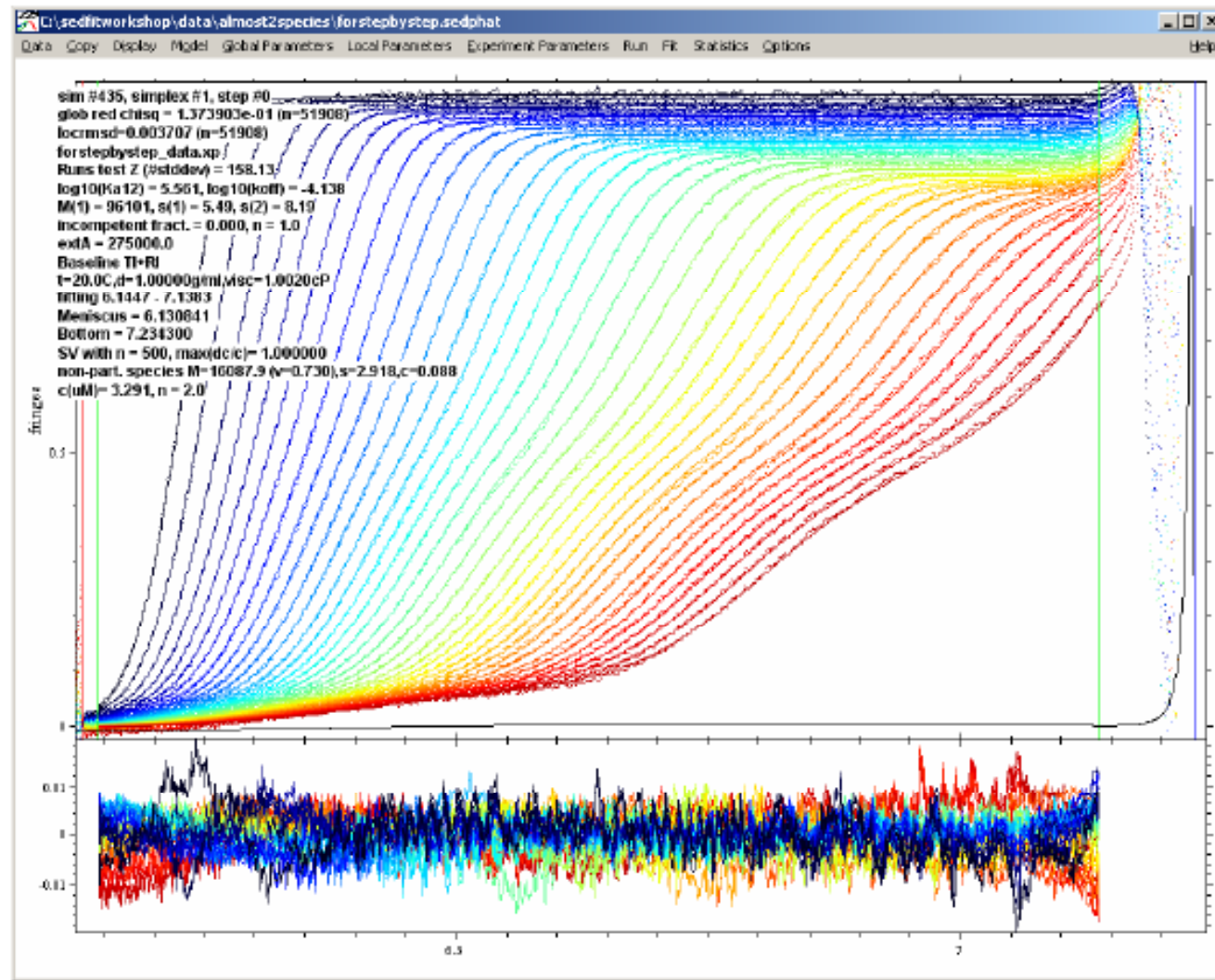
Again, click on the “Simplex” menu function. This will toggle off Marquardt-Levenberg and switch on the Simplex algorithm



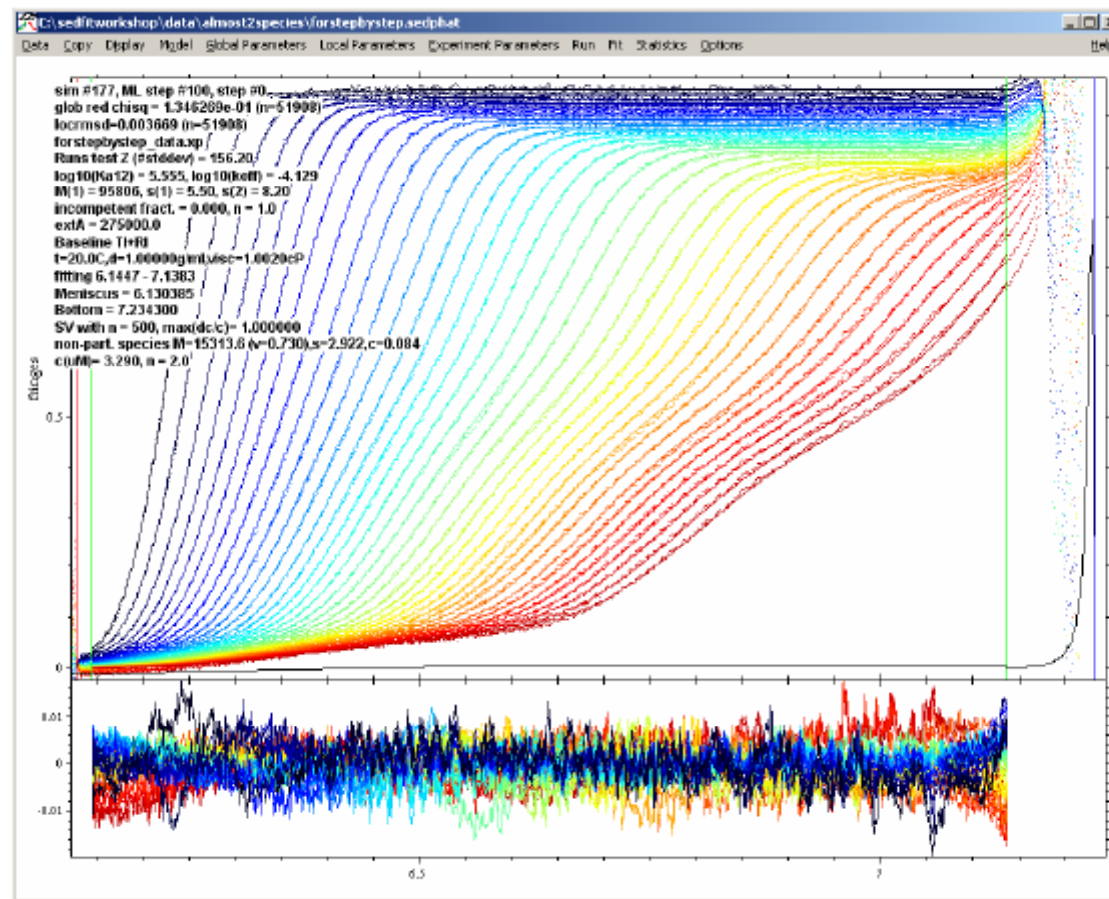
Then do a FIT

I find this still improves the fit slightly. After convergence gets very slow, I stopped it again and restarted the FIT. The reason for this is to take advantage of the random seeding of points each time you start a Simplex fit. This helped to get the rmsd further down. I repeat this procedure, but the second time it didn't seem to help.

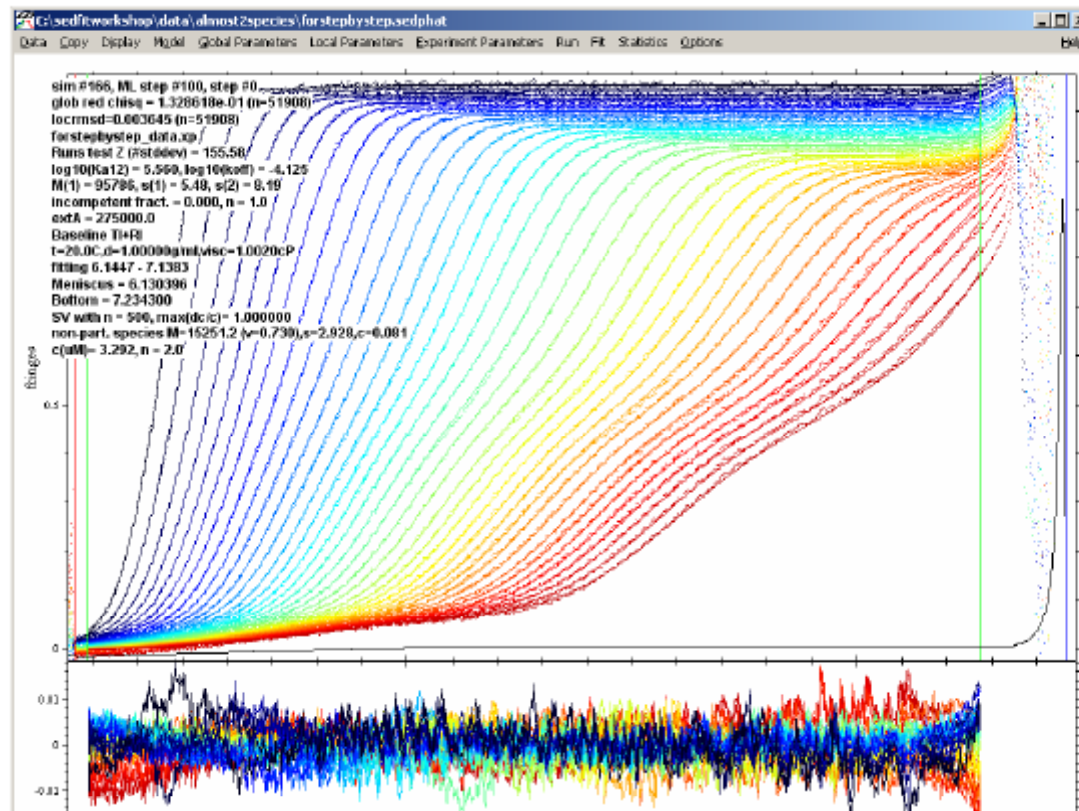
Save the configuration (control-U).



Then I switched back to the Marquardt-Levenberg method and tried again that way.
 This is the result: ***Further Improvement!!*** This is obviously a very badly shaped error surface, with a rather shallow valley that has many small ripples.



Same thing, again, after another round of alternating Simplex and Marquardt-Levenberg fits.



We could proceed in this way until we don't see any further improvement with any method. Another option would be to use the Simulated Annealing method, which I want to demonstrate in the following.

Simulated Annealing is better for error surfaces with many local minima, but more time-consuming. We could have used the Simulated Annealing method right from the beginning, but I didn't know yet that the error surface was so badly shaped.

At this point, let's float two more parameters that potentially could play a role in this fit:

in the **global parameters**:
switch on the incompetent
fraction

using only one experiment from a single concentration is not really very good to determine this parameter (because it likely will be correlated with the K_d), but the fine-points of the boundary shape could carry some information about it.

A + A <=> [A2]

Monomer	Dimer	incompetent fraction	non-participating species
☒ vary local concs c(k ₂) (uM) 3.292e+0	☒ log(K _a) 5.56035	☒ fraction 0.00000	☒ M 15251.21
excess (xp1) 275000.0	☒ log(k ₋) -4.12462		☒ s 2.928
☒ M(1) 95785.8	☒ s (2) 8.19489		☐ vbar 0.73
☒ s(1) 5.48499	☐ log10([k _s /(mg/ml)] nonid sec) -2.04580		☒ c
☒ Mass Conservation	☐ with 2nd virial coeff. MB [ml] 3.0000		
	☐ with hydrodyn kc		

Cancel
OK

in the experiment parameters:
switch on floating of the
bottom.

even though we should not see back-
diffusion from the main species, the
contaminating smaller Mw species may
experience back-diffusion slightly
influencing our residuals close to the
right fitting limit.

Experimental Parameters

(1) INTERFERENCE data for SEDVELOCITY

C:\sedit\workshop\data\almost2species\forstepbystep_data.xp (00005.IP2 ...)

Comment: cm/pixel: 0.00072708933, Sample

☒ active

noise: 0.0100

Centerpiece: 2

Pathlength: 1.200000

Rotor type: 4

☐ no backdiffusion noise

☒ fit baseline

☒ fit RI Noise

☒ fit TI Noise

☒ Meniscus 6.1304

☒ Bottom 7.2343

☐ redirect men./bot

For Associating Systems:

☐ extinction coefficient A 2.75000.00

☐ extinction coefficient B 0.0000

☐ extinction coefficient C 0.0000

☐ partial boundary fitting

amin: 0.00

amax: 0.00

Buttons: Cancel, OK

accept the default
limits for meniscus and
bottom variation.

DATA #1: set range for meniscus variation

enter upper limit: 6.160396

DATA #1: set range for meniscus variation

enter lower limit: 6.100390

DATA #1: set range for bottom variation

enter upper limit: 7.284300

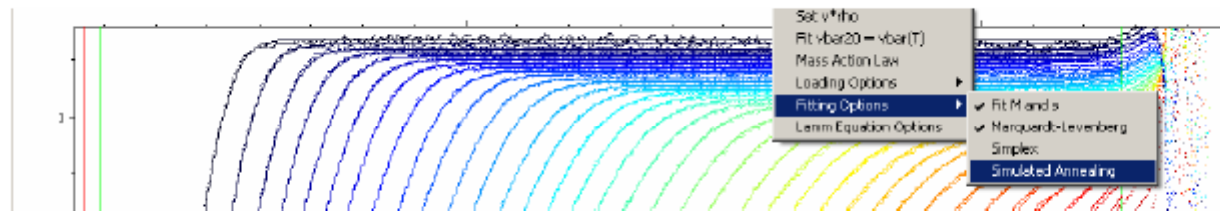
DATA #1: set range for bottom variation

enter lower limit: 7.184300

Buttons: OK, Cancel

before doing anything else, use SAVE CURRENT CONFIGURATION AS
and save under a different filename (I used 'forstepbystepSA').

Click on the 'Simulated Annealing' function. This will toggle it on (and the other algorithms off)



The following boxes will show up. Accept the default for all.

Then do a FIT
This will take a while.

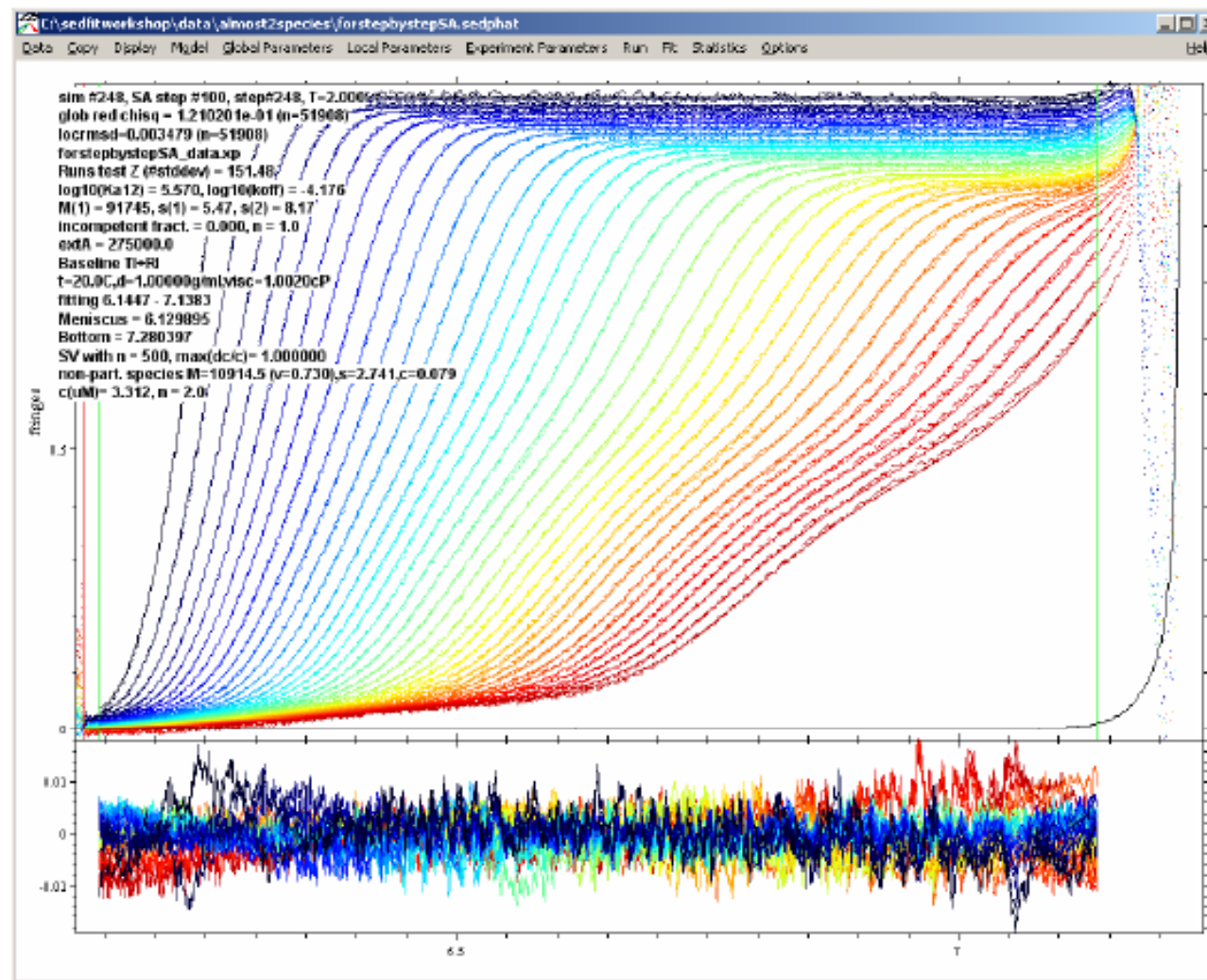
This will set the starting tolerance for accepting worse points during the procedure. A value of 2 means that points within a chisquare of twofold the starting chisquare will be tolerated.

This determines the 'temperature' schedule, i.e. the number of times when we reduce the threshold chisquare for rejecting worse points.

Since we will not expect any form of convergence, we have to predefine the number of iterations to be done at each temperature. You can get a feel for a ball-park number by looking at how many iterations the Simplex algorithm used previously.

We know that we likely will need another algorithm to home-in. The combination with a final round of Marquardt-Levenberg optimization seems to be reasonable.

This is where it ended up with SA – quite a bit better still. In fact, this is now better than the original $c(s)$ fit we started with. We should continue searching with alternate Simplex, Simulated Annealing and/or Marquardt-Levenberg steps for better fits.



Finally, after a second trial with Simulated Annealing, no improvement was possible with Simplex and Marquardt-Levenberg. This is the best fit I found. Unfortunately, there's no guarantee that this is really the global optimum.

