



Mathematical Best Fit Algorithm for Energy Distributions in VO Context

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Project Setup



Name

Mathematical fitting algorithms for Energy Distributions in VO context

Aim

To implement an algorithm that looks for the theoretical model that best fits a given Spectral Energy Distribution (SED)

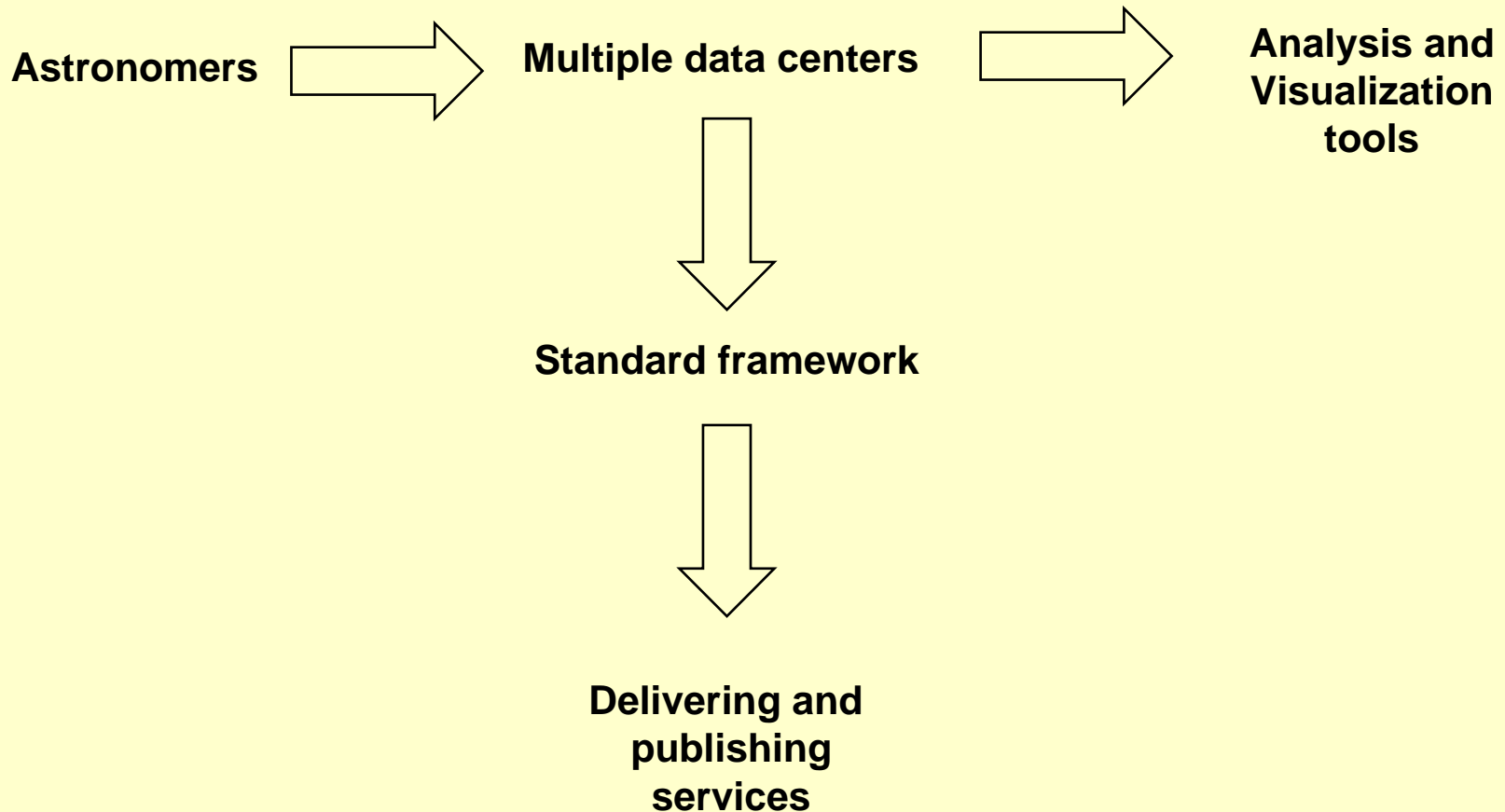
Tutors

**Pedro Osuna (ESA)
Jesus Salgado (INSA)**

Virtual Observatory



- Its aim is to offer seamless access to astronomical data worldwide





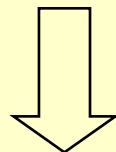
Virtual Observatory (cont.)

- In order to fulfill this aim the International Virtual Observatory Alliance (IVOA) was created in June 2002

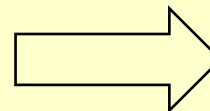
- **Data**
 - management
 - analysis
 - distribution
 - interoperability



- **Data Access Layer (DAL)**
 - working group within IVOA
 - standards for remote observational data access



Simple Spectral Access Protocol (SSAP)



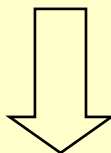
To define a uniform interface to spectral data



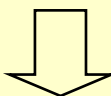
Virtual Observatory (cont.)

- **Theory Group**

- interest group within the IVOA
- formed en January 2004
- define requirements needed for a full interoperability between observations and models

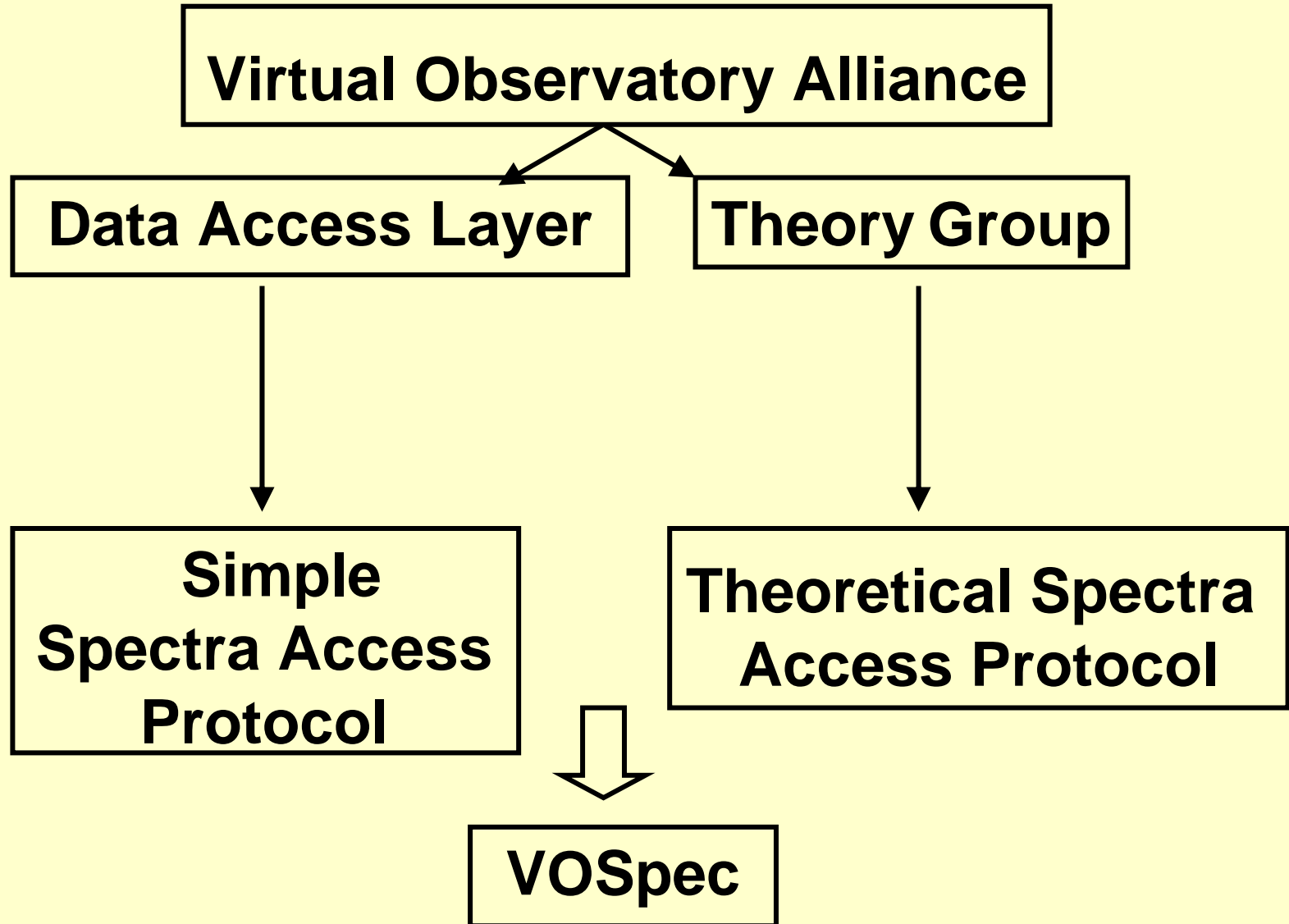


Theoretical Spectrum Access Protocol (TSAP)



Incorporate theoretical models to the VO

Project environment



VOSpec

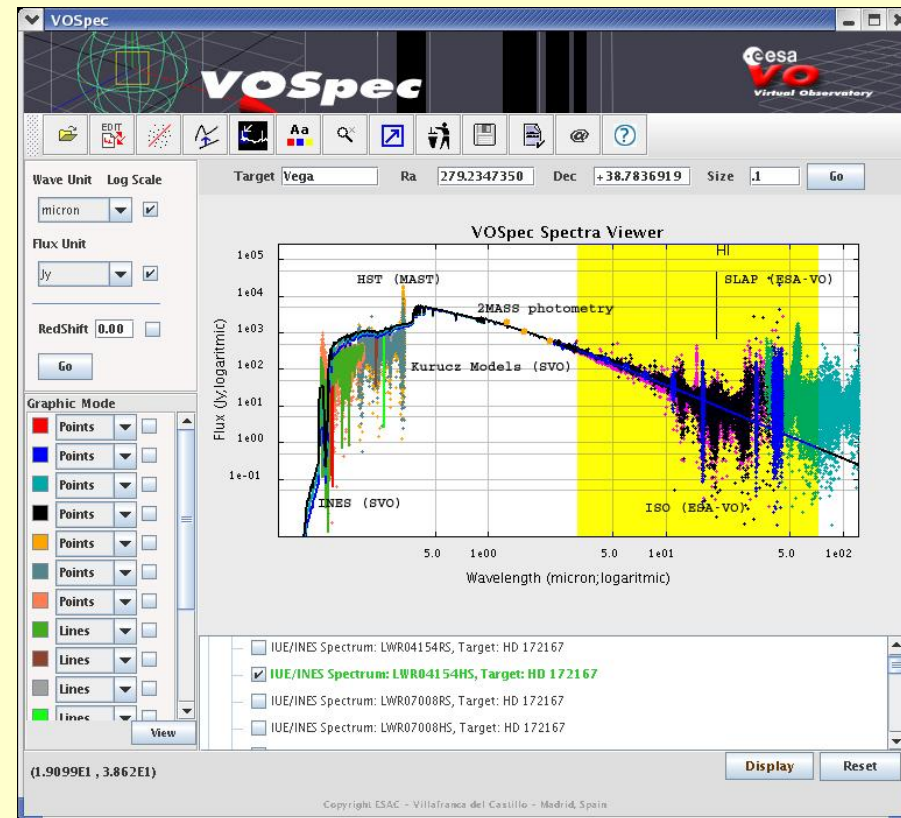


❑ Tool for handling

- VO compliant spectra through **SSAP**
- Theoretical models through **TSAP**

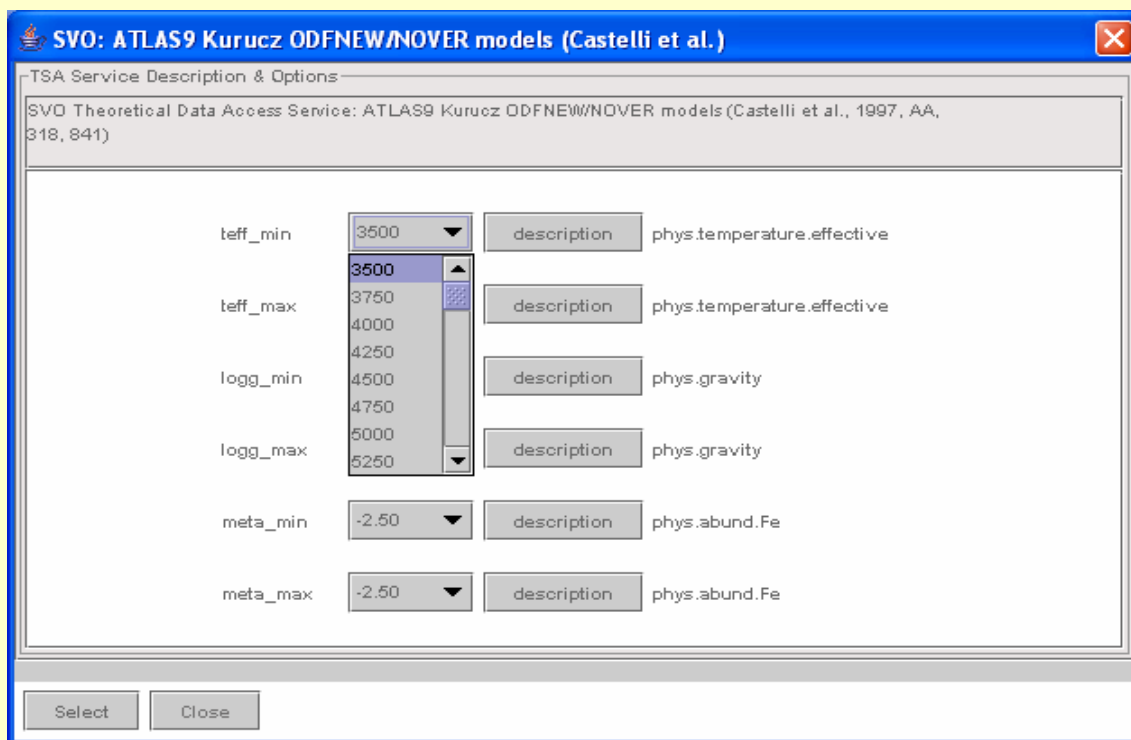
❑ Handling spectra with VOSpec:

- Display and superimpose spectra
- Automatic unit conversion through dimensional analysis
- Multi-wavelength analysis
- Polynomial/Black Body/Gaussian fitting



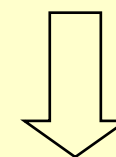


Why Best Fit?



Currently to find the model that best fits a given SED:

- Choose parameters manually
- Inspect result visually (“Chi-by-eye”)
- Modify parameters manually
- Reinspect visually
- Loop.....



New approach

Best Fit finds the model automatically

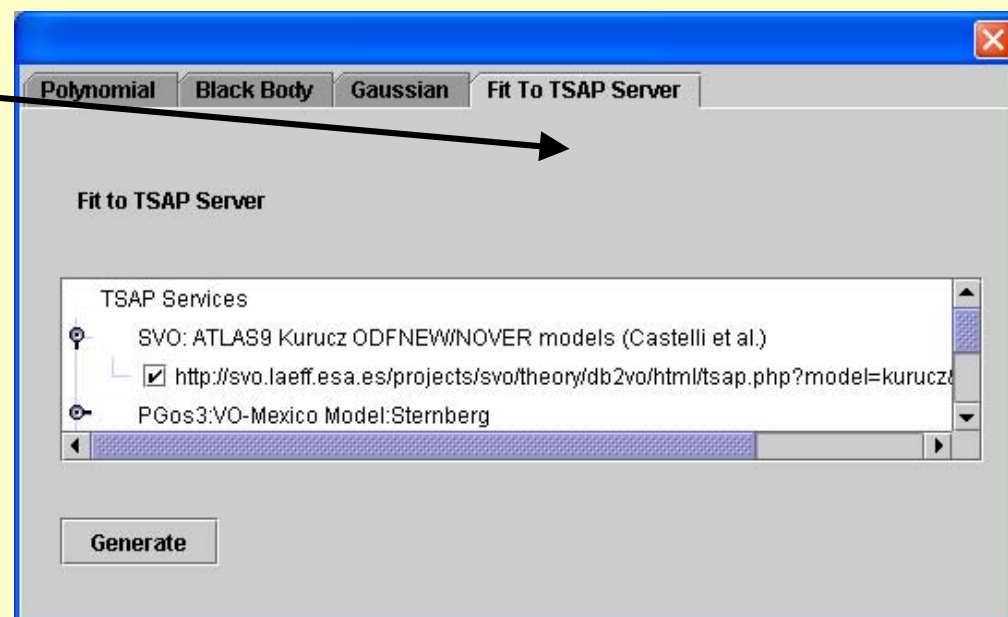
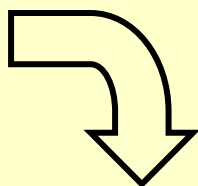


Best Fit brief description

❑ It is going to be integrated as a new fitting utility in **VOSpec**

❑ **Java** language

❑ **Aim**



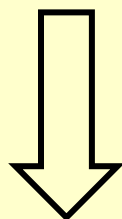
**Find the theoretical model
that best fits the SED**



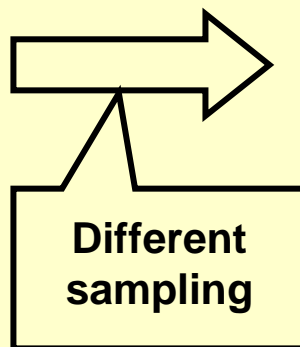
Best Fit brief description (cont.)

□ To fulfill this aim:

- Chi-square function measures the agreement between real and theoretical spectrum
- The best fit spectrum is the one that achieves a minimum in the Chi-square function
- Problem in minimization in many dimensions

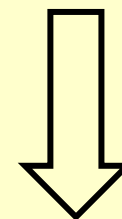


Levenberg-Marquardt method



$$\chi^2 = \sum_{i=1}^N \left(\frac{y_i - y(x_i; p_1, \dots, p_M)}{\sigma_i} \right)^2$$

real theoretical



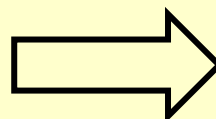
**SED
CUBIC SPLINE interpolation**



Levenberg-Marquardt method

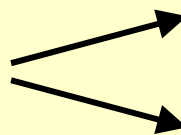
❑ Iterative method

❑ Standard technique for nonlinear least-square problems



Modified for discrete case

❑ Combination of



Steepest descent method

Gauss-Newton method

❑ **Steepest descent method**

• Initial estimate $\Rightarrow \mathbf{x}_0$

• Successive estimate $\Rightarrow \mathbf{x}_{i+1} = \mathbf{x}_i - c \cdot \nabla \chi^2(\mathbf{x}_i)$

• Iterations stop when $\nabla \chi^2(\mathbf{x}_i)$ is sufficiently small

• Hard to determine the step size

• Slow, but guaranteed convergence



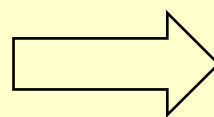
Levenberg-Marquardt method (cont.)

□ Gauss-Newton method

- Taylor expansion

$$\chi^2(\mathbf{x} + \Delta\mathbf{x}) = \chi^2(\mathbf{x}) + \nabla \chi^2(\mathbf{x}) \cdot \Delta\mathbf{x} + \frac{1}{2} \cdot \Delta\mathbf{x}^T \cdot \mathbf{H}(\mathbf{x}) \cdot \Delta\mathbf{x}$$

- Set the derivative of the previous expression equal to 0



Iterative equation

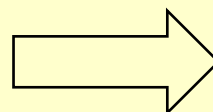
$$\mathbf{x}_{i+1} = \mathbf{x}_i - \mathbf{H}^{-1} \cdot \nabla \chi^2(\mathbf{x}_i)$$

- Quadratic convergence
- If the initial value is too far from the minimum convergence can fail



Levenberg-Marquardt method (cont.)

□ Steepest descent and Gauss-Newton methods are complementary in the advantages they provide



Levenberg Marquardt Method

$$\mathbf{x}_{i+1} = \mathbf{x}_i - c \cdot \nabla \chi^2(\mathbf{x}_i)$$

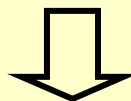
$$\mathbf{x}_{i+1} = \mathbf{x}_i - \mathbf{H}^{-1} \cdot \nabla \chi^2(\mathbf{x}_i)$$

$$\mathbf{x}_{i+1} = \mathbf{x}_i - (\mathbf{H} + \lambda \mathbf{I})^{-1} \cdot \nabla \chi^2(\mathbf{x}_i)$$



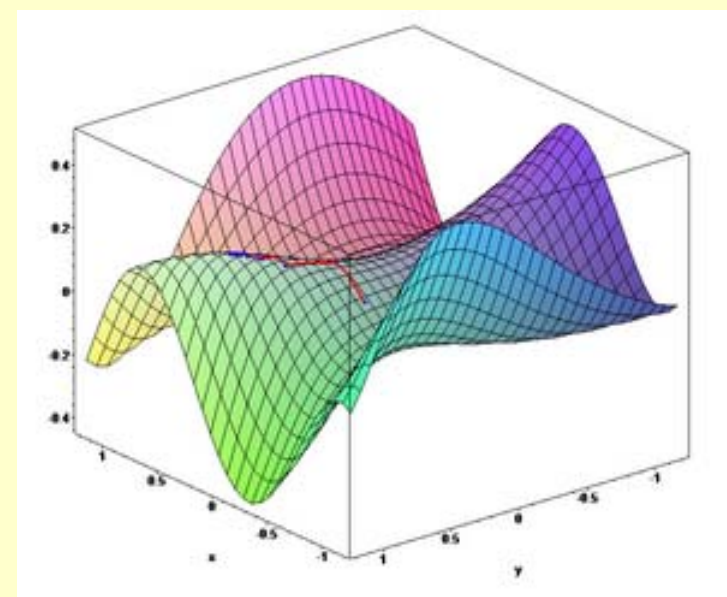
Levenberg-Marquardt method (cont.)

- **Far from the minimum**
 - Linear approximation with steepest descent method to the closest minimum
 - Adds a measure of the step in the steepest descent method



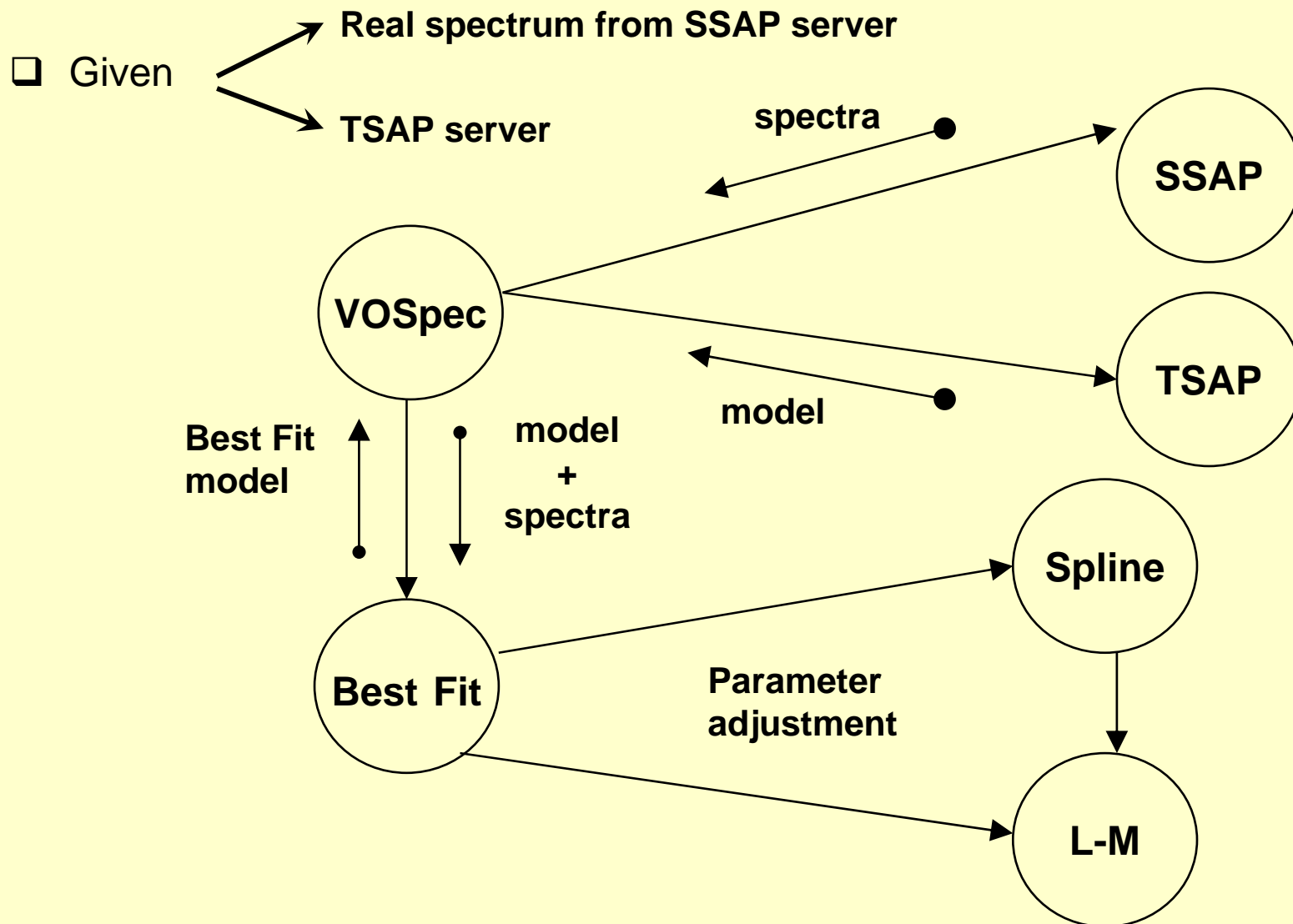
Close to the minimum
Gauss-Newton convergence

- **L-M method works well in practice**





Best Fit diagram





Conclusions

- IVOA standards are nearly mature enough to do real science
- BF algorithm implementation using TSAP protocol increases VO functionality
- Nonlinear least-square methods (LM) are very appropriate for the implementation of the algorithm
- Very easy integration with ESA VOSpec tool