

A short course about fitting models with the `scipy.optimize` module

Ariel Rokem¹

¹ The University of Washington eScience Institute

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Software

- [Review](#) ↗
- [Repository](#) ↗
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Description

Fitting models and testing the match of the models to the measured data is a fundamental activity in many fields of science. This short (approximately 3-hour) course (available at: <https://github.com/arokem/scipy-optimize>) aims to teach participants to use the Scipy library's `optimize` module to fit models to data (Jones et al. 2001–2001--). Using data from a psychology experiment (Rokem and Landau 2016) as an example, the course motivates the use of explicit mathematical models to explain and predict data and compares linear models and non-linear models. The core of the lesson focuses on fitting a curve with the `curve_fit` function. The course also introduces the idea of model comparison with cross-validation for evaluation and selection between non-nested non-linear models.

Statement of need

Model fitting is useful in many different fields of research, but optimization for model fitting is not a topic that is usually covered in introductory statistics or computing classes in many fields (e.g., psychology). This course fills an existing need for hands-on curriculum that goes beyond the topics taught in introductory computing workshops, such as ‘Software Carpentry’, providing material for follow-up workshops on advanced/intermediate topics. The target audience for this course are researchers or students with some programming knowledge (e.g., having participated in a ‘Software Carpentry’ workshop beforehand).

Learning objectives

In addition to these general objectives of this lesson, specific learning objectives are defined for each part of the lesson:

Part 1:

- Learners can define what a model is.
- Learners can define model parameters and model fitting.
- Learners can restate the benefits of modeling
- Learners can explain the utility of modeling applied to their data.

Part 2:

- Learners can identify a linear model.
- Learners can use `numpy` to fit a linear model to data
- Learners can evaluate a model using model residuals.

Part 3:

- Learners can identify a nonlinear model, and discuss the differences between linear and nonlinear models
- Learners can use `scipy.optimize` to fit a nonlinear model to data.
- Learners can calculate and display model residuals for nonlinear models

Part 4:

- Learners can define and identify overfitting.
- Learners can implement split-half cross-validation to evaluate model error.

Usage

To use these instructional materials, it is recommended that the instructor type out the code in an interactive environment, such as a [Jupyter](#) notebook, while learners follow along on their own machines. For this purpose, prerequisites and setup instructions are provided on the [first page](#) of the lesson.

Contributing

Contributing to the module

Contributions, corrections and improvements are welcome through a pull request against the repository at <https://github.com/arokem/scipy-optimize>

Reporting issues or problems with the module and seeking support

Issues and problems with the module, as well as requests for support, should be submitted using the GitHub repo Issues page at: <https://github.com/arokem/scipy-optimize/issues>

Content, instructional design, and experience of use in teaching and learning situations

This lesson, initially a blog post produced as a final exercise for Software Carpentry Instructor Training (<http://arokem.github.io/2014-08-12-learn-optimization.html>), was first used as lesson as a contribution to a series of half-day workshops organized by the Lab for Data Intensive Biology at UC Davis (and described [here](#))

A video of this example of instruction using this lesson is available on [YouTube](#)

References

Jones, Eric, Travis Oliphant, Pearu Peterson, and others. 2001–2001--.. “SciPy: Open Source Scientific Tools for Python.” <http://www.scipy.org/>.

Rokem, Ariel, and Ayelet Nina Landau. 2016. “The Interaction of Orientation-Specific Surround Suppression and Visual-Spatial Attention.” *bioRxiv*. <https://doi.org/10.1101/091553>.