# Parameter Minimization using the

# Nelder-Mead algorithm

Do some fitting

### 1.1 Introduction

The Nelder-Mead plugin is used to fit a SBML models parameters to experimental data.

The plugin has numerous properties to allow the user full control over the internal fitting engine, as well as access to generated fitted data after a minimization session. In addition, various statistical properties, such as standardized residuals, Q-Q data, ChiSquare and reduced ChiSquare are made accessible to the user. The resulting parameter values do also come with estimated confidence limits.

The current implementation is based on the Nelder-Mead C implementation by Michael F. Hutt<sup>1</sup>.

Plugin properties are documented in the next section.

1

### 1.2 Plugin Properties

Available properties in the Nelder-Mead plugin are listed in the table below.

Property Name	Data Type	Default Value	Description
SBML	string	N/A	SBML document as a string. Model to be used in the
			fitting.
ExperimentalData	telluriumData	N/A	Input data.
${\bf FittedData}$	telluriumData	N/A	Output data.
Input Parameter List	listOfProperties	N/A	Parameters to fit.
${\bf Output Parameter List}$	listOfProperties	N/A	List of fitted parameters.
Experimental-	stringList	N/A	Species selection list for experimental data.
DataSelectionList			
${\bf Fitted Data Selection Lis}$	st stringList	N/A	Selection list for model data.
Norm	double	N/A	Norm of fitting. An estimate of goodness of fit.
Norms	telluriumData	N/A	The norm is calculated throughout a fitting session.
			Each Norm value is stored in the Norms (read-only)
			property.
ConfidenceLimits	listOfProperties	N/A	Confidence limits for each fitted parameter. The confi-
			dence limits are calculated at a $95\%$ confidence level.
Hessian	matrix	N/A	Hessian matrix. The Hessian is calculated using approx-
			imation at a found parameter minimum.

CovarianceMatrix	matrix	N/A	Covariance matrix. Calculated as the inverse of the Hes-
			sian.
Residuals	telluriumData	N/A	Residuals data.
${\bf Standardized Residuals}$	telluriumData	N/A	Standardized Residuals.
NormalProb-	telluriumData	N/A	Normal Probability of Residuals.
${\it ability} Of Residuals$			
ChiSquare	double	N/A	The ChiSquare at the minimum.
ReducedChiSquare	double	N/A	The Reduced ChiSquare at the minimum.
${\bf Status Message}$	string	N/A	Message from the internal fitting engine, communicating
			the status of the obtained fit (Currently not used).
NrOfIter	int	N/A	Number of (internal outer loop) iterations.
NrOfFuncIter	int	N/A	Number of objective function iterations.

The following properties are used internally by the fitting engine. They are pre-set with default values.

Depending on the minimization problem at hand, they may need to be tweaked.

Epsilon	double	1.e-6	Convergence tolerance.
Scale	double	1	Scaling of vertices.
${\bf MaxNrOfIterations}$	int	1000	Maximum number of iterations.
Alpha	double	1.	Reflection coefficient.

Beta	double	0.5	Contraction coefficient.
Gamma	double	1	Expansion coefficient.

Table 1.1: Plugin Properties

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### 1.3 Plugin Events

The Nelder-Mead plugin are using all of a plugins available plugin events, i.e. the *PluginStarted*, *PluginProgress* and the *PluginFinished* events.

The available data variables for each event are internally treated as *pass trough* variables, so any data, for any of the events, assigned prior to the plugins execute function (in the assingOn.. family of functions), can be retrieved unmodified in the corresponding event function.

Event	Arguments	Purpose and argument types
PluginStarted	void*, void*	Signal to application that the plugin has started.
		Both parameters are pass trough parameters and
		are unused internally by the plugin.
PluginProgress	void*, void*	Communicating progress of fitting. Both param-
		eters are <i>pass trough</i> parameters and are unused internally by the plugin.
PluginFinished	void* void*	Signals to application that execution of the plugin
r rugiiir iiiisiica	void , void	has finished. Both parameters are pass trough pa-
		rameters and are unused internally by the plugin.

Table 1.2: Plugin Events

### 1.4 Python example

The following Python script illustrate how the plugin can be used.

```
1
  from telplugins import *
3
  # Load Plugins
4
  chiPlugin
                 = Plugin("tel_chisquare")
                 = Plugin("tel_nelder_mead")
5
                 = Plugin("tel_test_model")
   modelPlugin
7
   addNoisePlugin = Plugin("tel_add_noise")
8
9
      10
      def myEvent(dummy): #We are not capturing any data from the plugin, so
11
         just pass a dummy
          print 'Iteration, Norm = ' + 'nm.getProperty("NrOfIter")' + ', ' +
12
              'nm.getProperty("Norm")'
13
14
      #Setup progress event function
15
      progressEvent = NotifyEventEx(myEvent)
16
      assignOnProgressEvent(nm.plugin, progressEvent)
```

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```
#-----
17
18
19
       #Create model data, with and without noise using the test_model plugin
20
       modelPlugin.execute()
21
22
       #Setup plugin properties.
23
       nm.SBML
                           = modelPlugin.Model
24
       nm.ExperimentalData = modelPlugin.TestDataWithNoise
25
26
       # Add the parameters that we're going to fit and an initial 'start'
          value
27
       nm.setProperty("InputParameterList", ["k1", .3])
       nm.setProperty("FittedDataSelectionList", "[S1] [S2]")
28
29
       nm.setProperty("ExperimentalDataSelectionList", "[S1] [S2]")
30
31
       # Start minimization
32
       nm.execute()
33
34
       print 'Minimization finished. \n==== Result ===='
35
36
       print 'Hessian Matrix'
37
       print nm.getProperty("Hessian")
38
39
       print 'Covariance Matrix'
40
       print nm.getProperty("CovarianceMatrix")
41
                                      + 'nm.getProperty("ChiSquare")'
       print 'ChiSquare = '
42
       print 'Reduced ChiSquare = ' + 'nm.getProperty("ReducedChiSquare")'
43
44
45
       #This is a list of parameters
46
       parameters = tpc.getPluginProperty (nm.plugin, "OutputParameterList")
47
       confLimits = tpc.getPluginProperty (nm.plugin, "ConfidenceLimits")
48
49
       #Iterate trough list of parameters and confidence limits
50
       para = getFirstProperty(parameters)
51
       limit = getFirstProperty(confLimits)
       while para and limit:
52
           print getPropertyName(para) + ' = ' + 'getPropertyValue(para)' + '
53
              +/- ' + 'getPropertyValue(limit)'
           para = getNextProperty(parameters)
54
55
           limit = getNextProperty(confLimits)
56
57
58
       # Get the fitted and residual data
       fittedData = nm.getProperty ("FittedData").toNumpy
59
60
       residuals = nm.getProperty ("Residuals").toNumpy
61
62
       # Get the experimental data as a numpy array
       experimentalData = modelPlugin.TestDataWithNoise.toNumpy
63
64
65
       telplugins.plot(fittedData
                                         [:,[0,1]], "blue", "-",
          S1 Fitted")
                                          [:,[0,2]], "blue", "-",
66
       telplugins.plot(fittedData
          S2 Fitted")
```

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```
[:,[0,1]], "blue", "None", "x",
67
       telplugins.plot(residuals
          S1 Residual")
68
       telplugins.plot(residuals
                                           [:,[0,2]], "red", "None", "x",
          S2 Residual")
69
       telplugins.plot(experimentalData
                                           [:,[0,1]], "red", "",
          S1 Data")
                                           [:,[0,2]], "blue", "",
70
       telplugins.plot(experimentalData
          S2 Data")
71
       telplugins.plt.show()
72
73
       \#Finally, view the manual and version
74
       nm.viewManual()
       print 'Plugin version: ' + 'nm.getVersion()'
75
76
77
   except Exception as e:
       print 'Problem.. ' + 'e'
78
```

Listing 1.1: Minimization example.

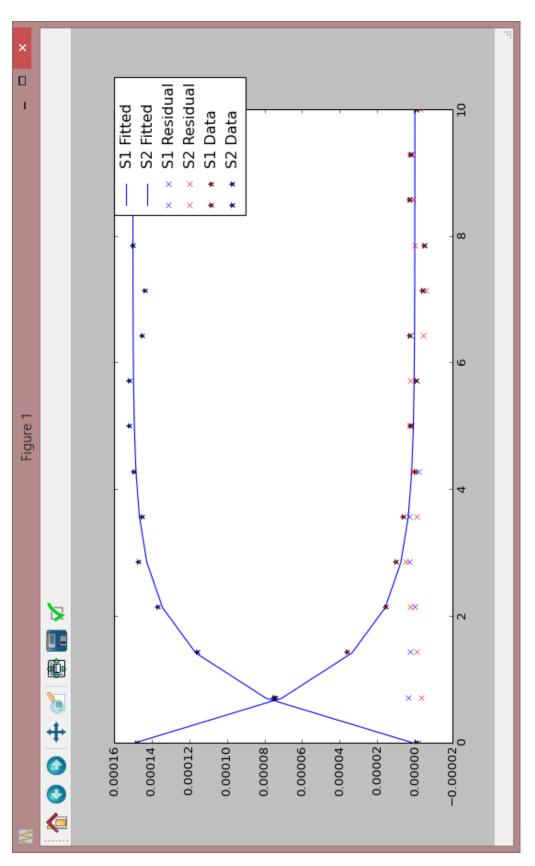


Figure 1.1: Typical output for the example script above.