

PS1_pyglotaran

March 27, 2021

1 Notes K-matrix

- $K_{ij} \iff$ energy flow from state j to state i

1.1 parallel (DAS)

Energy leaves system directly

Full matrix \iff Reduced matrix:

- $K_{ij} = 0 \forall i \neq j$

1.2 sequential (EAS)

Energy flows from one compartment i to the next $i + 1$ until it leaves the system (K_{nn})

Assume \mathbf{K} is $n \times n$ and sequence is in order

Full matrix:

- $K_{i+1i} = -K_{ii} \forall i < n \wedge K_{nn} \neq 0$

Reduced matrix:

- $K_{i+1i}^{Reduced} = K_{i+1i}^{Full} \wedge K_{ii} = 0 \forall i < n \wedge K_{nn} \neq 0$

1.3 General

Full matrix:

- $K_{ii} \iff$ - sum flows out of the compartment i (to other compartments or out of the system)
- $K_{ij} \forall i \neq j \iff$ flow from compartment j to compartment i

Full matrix:

- $\left(\sum_{k \neq i}^n K_{ki}\right) - K_{ii}^{Full} \iff$ flow of the compartment i out of the system
- $K_{ij}^{Reduced} = K_{ij}^{Full} \forall i \neq j \iff$ flow from compartment j to compartment i

2 cyano-PS1 case study

As in the lecture note we first try a global analysis

2.1 Global analysis

Assumption the process can be described by a simple sequential model.

2.1.1 K-matrix sequential

The full K-matrix for a 4 compartment sequential scheme has the form of \mathbf{K}^{Full}

$$\mathbf{K}^{Full} = \begin{bmatrix} -k_{21} & & & & \\ k_{21} & -k_{32} & & & \\ & k_{32} & -k_{43} & & \\ & & k_{43} & k_{4Out} & \\ & & & & \end{bmatrix}$$

Since both glotaran and pyglotaran use a reduced form of the K-matrix, it needs to be transformed

$$\mathbf{K}^{Reduced} = \begin{bmatrix} k_{21} & & & & \\ & k_{32} & & & \\ & & k_{43} & & \\ & & & k_{4Out} & \\ & & & & \end{bmatrix}$$

2.1.2 Notebook helper function for pretty output

```
[1]: from IPython.display import Markdown
from IPython.display import display

def print_md(markdown_printable):
    """Pretty render Markdown."""
    display(Markdown(str(markdown_printable)))

def print_yaml_file(file_path):
    """Pretty render yaml file."""
    with open(file_path) as f:
        print_md(f"`yaml\n{f.read()}\n`")
```

2.1.3 Plotting functions (pyglotaran_extras + matplotlib)

```
[2]: from pyglotaran_extras.plotting.plot_svd import plot_svd
from pyglotaran_extras.plotting.plot_overview import plot_overview
import matplotlib.pyplot as plt
plt.rcParams["figure.figsize"] = (21,14)
```

2.1.4 Analysis functions

```
[3]: from glotaran.analysis.optimize import optimize
from glotaran.analysis.scheme import Scheme
from glotaran import read_model_from_yaml_file
from glotaran import read_parameters_from_yaml_file
```

```
from glotaran.io import read_data_file
```

2.1.5 Read data

```
[4]: dataset = read_data_file("streakdata.ascii")
dataset
```

```
[4]: <xarray.Dataset>
Dimensions:                (left_singular_value_index: 49,
right_singular_value_index: 49, singular_value_index: 49, spectral: 49, time:
923)
Coordinates:
  * time                    (time) float64 -101.0 -100.8 ... 99.24 99.46
  * spectral                (spectral) float64 626.1 629.5 ... 785.3 788.7
Dimensions without coordinates: left_singular_value_index,
right_singular_value_index, singular_value_index
Data variables:
  data                      (time, spectral) float64 50.75 59.21 ... 113.1
  data_left_singular_vectors (time, left_singular_value_index) float64 -0...
  data_singular_values       (singular_value_index) float64 2.802e+05 ...
  data_right_singular_vectors (right_singular_value_index, spectral) float64
...
```

2.1.6 Load model and parameters

```
[5]: global_model = read_model_from_yaml_file("models/global-model.yml")
global_parameters = read_parameters_from_yaml_file("models/global-parameters.
→yml")
print_md(global_model.validate(parameters=global_parameters))
```

Your model is valid.

```
[6]: print_md(global_model)
```

3 Model

Type: kinetic-spectrum

3.1 Initial Concentration

- **input1:**
 - *Label:* input1
 - *Compartments:* ['s1', 's2', 's3', 's4']
 - *Parameters:* [input.1, input.0, input.0]
 - *Exclude From Normalize:* []

3.2 K Matrix

- **km1:**
 - *Label:* km1
 - *Matrix:*
 - * ('s2', 's1'): kinetic.1
 - * ('s3', 's2'): kinetic.2
 - * ('s4', 's3'): kinetic.3
 - * ('s4', 's4'): kinetic.4

3.3 Irf

- **irf1** (gaussian):
 - *Label:* irf1
 - *Type:* gaussian
 - *Center:* irf.center
 - *Width:* irf.width
 - *Normalize:* False
 - *Backsweep:* True
 - *Backsweep Period:* irf.backsweep

3.4 Dataset

- **dataset1:**
 - *Label:* dataset1
 - *Megacomplex:* ['mc1']
 - *Initial Concentration:* input1
 - *Irf:* irf1

3.5 Megacomplex

- **mc1:**
 - *Label:* mc1
 - *K Matrix:* ['km1']

3.5.1 Create sceme and optimize it

```
[7]: global_scheme = Scheme(global_model,global_parameters,{'dataset1':  
    ↳dataset},non_negative_least_squares=True)  
global_result = optimize(global_scheme)
```

Iteration	Total nfev	Cost	Cost reduction	Step norm
Optimality				
0	1	4.7450e+07		
2.84e+10				
1	2	4.3767e+07	3.68e+06	1.55e-01
5.88e+08				
2	3	4.3750e+07	1.73e+04	1.16e-02
1.63e+08				

3	4	4.3750e+07	6.34e+01	3.25e-04
2.50e+07				
4	5	4.3750e+07	6.14e-01	2.41e-05
2.84e+06				
5	6	4.3750e+07	8.45e-03	2.04e-06
4.06e+05				

`ftol` termination condition is satisfied.
Function evaluations 6, initial cost 4.7450e+07, final cost 4.3750e+07, first-order optimality 4.06e+05.

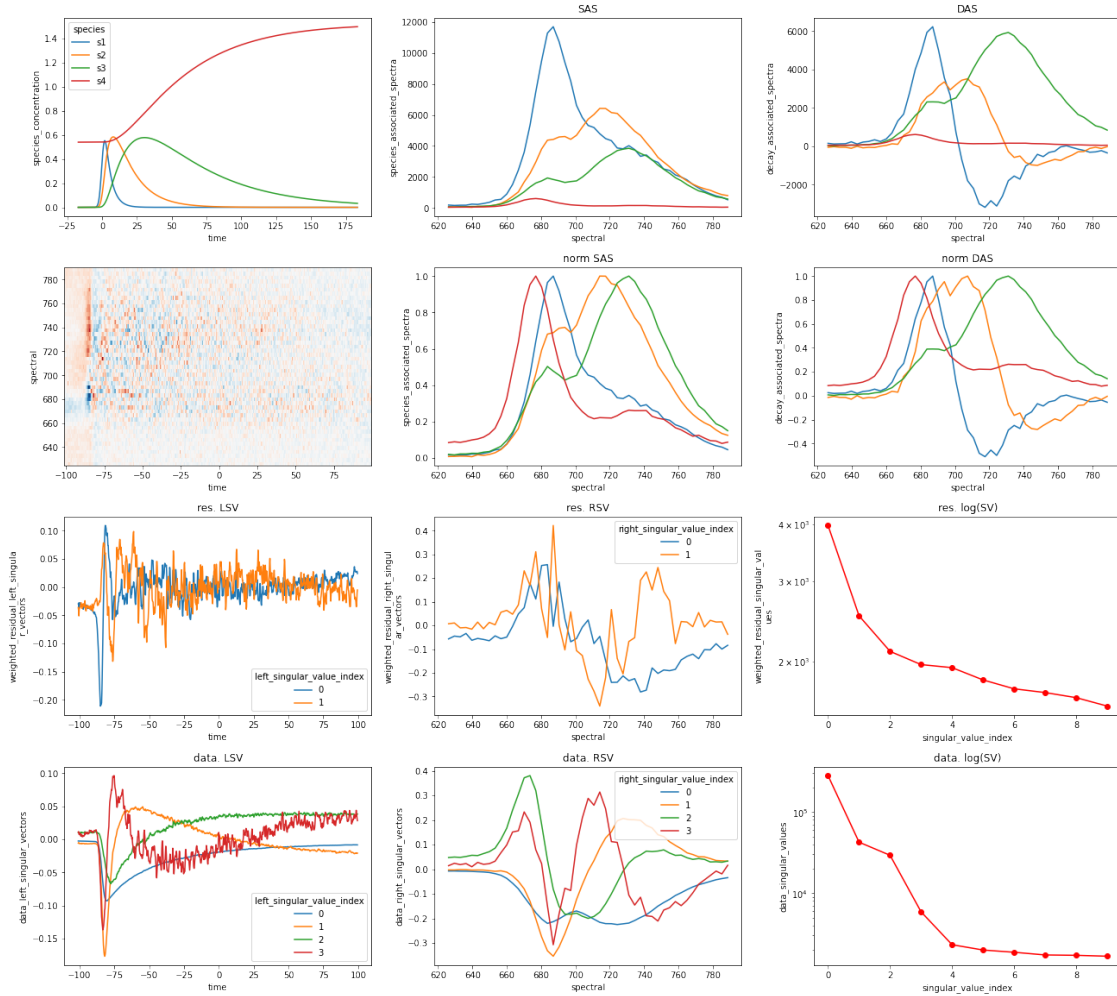
```
[8]: global_result.data['dataset1']
```

```
[8]: <xarray.Dataset>
Dimensions:                                (clp_label: 4, component: 4,
from_species: 4, left_singular_value_index: 49, right_singular_value_index: 49,
singular_value_index: 49, species: 4, spectral: 49, time: 923, to_species: 4)
Coordinates:
  * time                                     (time) float64 -101.0 ... 99.46
  * spectral                               (spectral) float64 626.1 ... 788.7
  * clp_label                             (clp_label) <U2 's1' 's2' ... 's4'
  * species                               (species) <U2 's1' 's2' 's3' 's4'
    rate                                  (component) float64 -0.225 ... ..
    lifetime                             (component) float64 -4.445 ... ..
  * to_species                             (to_species) <U2 's1' ... 's4'
  * from_species                           (from_species) <U2 's1' ... 's4'
Dimensions without coordinates: component, left_singular_value_index,
right_singular_value_index, singular_value_index
Data variables: (12/24)
  data                                     (time, spectral) float64 50.75 ...
  data_left_singular_vectors              (time, left_singular_value_index)
float64 ...
  data_singular_values                    (singular_value_index) float64 ...
  data_right_singular_vectors              (spectral,
right_singular_value_index) float64 ...
  matrix                                  (time, clp_label) float64 9.683...
  clp                                      (spectral, clp_label) float64 1...
  ...                                     ...
  a_matrix                                (component, species) float64 1...
  k_matrix                                (to_species, from_species) float64
...
  k_matrix_reduced                        (to_species, from_species) float64
...
  irf_center                             float64 -83.85
  irf_width                              float64 1.61
  irf                                     (time) float64 2.691e-25 ... 0.0
Attributes:
  root_mean_square_error:                 43.98496106181416
```

weighted_root_mean_square_error: 43.98496106181416

3.5.2 Result plots

```
[9]: fig = plot_overview(global_result.data['dataset1'],linlog=False)
```



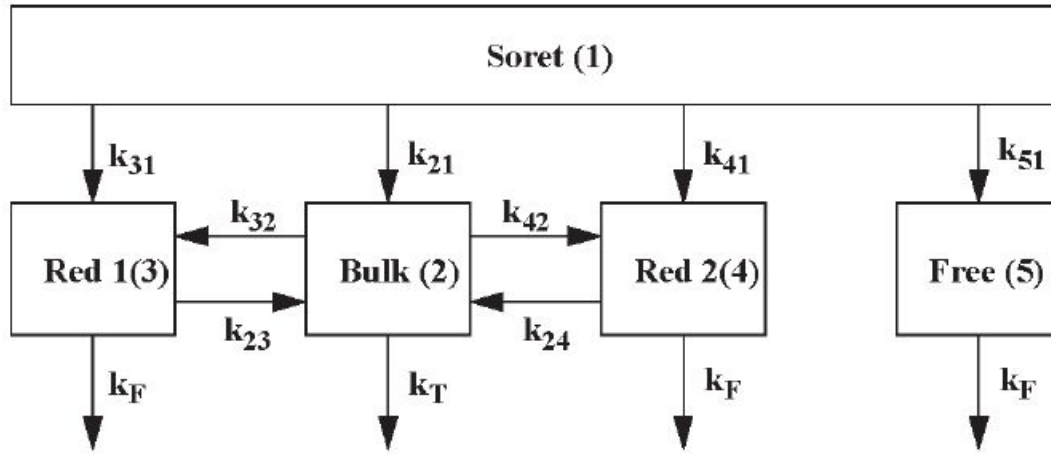
```
[10]: print_md(global_result.optimized_parameters)
```

- **None:**
 - **input:**
 - * **1:** Value: 1.0, StdErr: 0.0, Min: -inf, Max: inf, Vary: False, Non-Negative: False, Expr: None
 - * **0:** Value: 0.0, StdErr: 0.0, Min: -inf, Max: inf, Vary: False, Non-Negative: False, Expr: None
 - **irf:**
 - * **center:** Value: -83.85334624849767, StdErr: 6.733909520056759e-05, Min: -inf, Max: inf, Vary: True, Non-Negative: False, Expr: None

- * **width**: *Value*: 1.6098575199546155, *StdErr*: 6.952805066648674e-05, *Min*: -inf, *Max*: inf, *Vary*: True, *Non-Negative*: False, *Expr*: None
- * **backsweep**: *Value*: 13200.0, *StdErr*: 0.0, *Min*: -inf, *Max*: inf, *Vary*: False, *Non-Negative*: False, *Expr*: None
- **kinetic**:
 - * **1**: *Value*: 0.22498267531276447, *StdErr*: 3.32612259772865e-05, *Min*: -inf, *Max*: inf, *Vary*: True, *Non-Negative*: False, *Expr*: None
 - * **2**: *Value*: 0.06806909027669798, *StdErr*: 1.1979156773469584e-05, *Min*: -inf, *Max*: inf, *Vary*: True, *Non-Negative*: False, *Expr*: None
 - * **3**: *Value*: 0.021225130267494157, *StdErr*: 1.186486010552218e-06, *Min*: -inf, *Max*: inf, *Vary*: True, *Non-Negative*: False, *Expr*: None
 - * **4**: *Value*: 0.0001596930041049832, *StdErr*: 2.2438163466850647e-08, *Min*: -inf, *Max*: inf, *Vary*: True, *Non-Negative*: False, *Expr*: None

3.6 Target analysis

Theoretical background from `compartmental_models.pdf`



Compartmental model describing the kinetics of cyanobacterial PS I core particles upon excitation at 400 nm.

Concentration vector

$$c(t) = [S(t) \ B(t) \ R_1(t) \ R_2(t) \ F(t)]^T$$

The concentration function correspond to

<i>Function name</i>	<i>S(t)</i>	<i>B(t)</i>	<i>R_1(t)</i>	<i>R_2(t)</i>	<i>F(t)</i>
Name in the schema	Soret	Bulk	Red1	Red2	Free

Differential equation

$$\frac{d}{dt}c(t) = \mathbf{K}c(t) + j(t)$$

Input function

$$j(t) = i(t) \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \end{bmatrix}^T$$

Transfer maxtrix \mathbf{K}

$$\mathbf{K} = \begin{bmatrix} -(k_{21} + k_{31} + k_{41} + k_{51}) & & & & \\ k_{21} & -(k_T + k_{32} + k_{42}) & k_{23} & k_{24} & \\ k_{31} & k_{32} & -(k_F + k_{23}) & & \\ k_{41} & k_{42} & & -(k_F + k_{24}) & \\ k_{51} & & & & k_F \end{bmatrix}$$

- k_T : effective rate constant of Chl, T stands for photochemical Trapping of the excitation energy
- k_F : natural decay rate of free Chl, F stands for Fluorescence

```
[11]: # Just a little helper to quickly make a pretty table
from tabulate import tabulate
headers = list(map(lambda x: f"***{x}***", ["Function name", "S(t)", "B(t)", "R_1(t)", "R_2(t)", "F(t)"]))
print(tabulate([["**Name in the schema**", "Soret", "Bulk", "Red1", "Red2", "Free"]], headers=headers, tablefmt="github"))
```

```
| ***Function name*** | ***S(t)*** | ***B(t)*** | ***R_1(t)*** | ***R_2(t)*** | ***F(t)*** |
|-----|-----|-----|-----|-----|
| **Name in the schema** | Soret | Bulk | Red1 | Red2 | Free |
```

3.6.1 Reduced K-matrix

$$\mathbf{K} = \begin{bmatrix} k_{21} & k_T & k_{23} & k_{24} \\ k_{31} & k_{32} & k_F & \\ k_{41} & k_{42} & & k_F \\ k_{51} & & & k_F \end{bmatrix}$$

3.6.2 Reduced K-matrix ignoring input to Red1 and Red2

$$\mathbf{K} = \begin{bmatrix} k_{21} & k_T & k_{23} & k_{24} \\ & k_{32} & k_F & \\ & k_{42} & & k_F \\ k_{51} & & & k_F \end{bmatrix}$$

With the rate relations:

- $k_{23} = a_1 k_{32}$
- $k_{24} = a_2 k_{42}$

3.6.3 target-model.yml

```
[12]: print_yaml_file("models/target-model.yml")

type: kinetic-spectrum

megacomplex:
  mc1:
    k_matrix: [km1]

k_matrix:
  km1:
    matrix:
      (s2, s1): kinetic.1
      (s5, s1): kinetic.2
      (s2, s2): kinetic.3
      (s3, s2): kinetic.4
      (s4, s2): kinetic.5
      # (s2, s3): {expr:scaling.a1*kinetic.4}
      # (s2, s3): $scaling.1 * $kinetic.4
      (s2, s3): kinetic.7
      (s3, s3): kinetic.6
      # (s2, s4): {expr:scaling.a2*kinetic.5}
      # (s2, s4): $scaling.2 * $kinetic.5
      (s2, s4): kinetic.8
      (s4, s4): kinetic.6
      (s5, s5): kinetic.6

spectral_constraints:
  - type: zero
    compartment: s1
    interval:
      - [100, 1000]
  - type: zero
    compartment: s3
    interval:
      - [100, 680]
  - type: zero
    compartment: s4
    interval:
      - [100, 690]

irf:
  irf1:
    type: gaussian
    center: irf.center
    width: irf.width
```

```

    backsweep: True
    backsweep_period: irf.backsweep

initial_concentration:
  input1:
    compartments: [s1, s2, s3, s4, s5]
    parameters: [
      input.1,
      input.0,
      input.0,
      input.0,
      input.0]

dataset:
  dataset1:
    initial_concentration: input1
    megacomplex: [mc1]
    irf: irf1

```

3.6.4 target-parameters

```
[13]: print_yaml_file("models/target-parameters.yml")
```

```

input:
  - ["1", 1]
  - ["0", 0]
  - {vary: false}

irf:
  - ["center", -84.0]
  - ["width", 1.6]
  - ["backsweep", 13800.0, {vary: false}]

kinetic:
  - [2, {vary: false}]
  - [0.1, {vary: false}]
  - 0.05
  - 0.1
  - 0.03
  - 0.00016
  - ["7", {expr: '$kinetic.4 * $scaling.1'}]
  - ["8", {expr: '$kinetic.5 * $scaling.2'}]
#   - {non-negative: true}

scaling:

```

- [1.87, {vary: false}]
- [1.37, {vary: false}]

```
[14]: target_model = read_model_from_yaml_file("models/target-model.yml")
      target_parameters = read_parameters_from_yaml_file("models/target-parameters.
      ↪yml")
      print_md(target_model.validate(parameters=target_parameters))
```

Your model is valid.

```
[15]: print_md(target_model)
```

4 Model

Type: kinetic-spectrum

4.1 Initial Concentration

- **input1:**
 - *Label:* input1
 - *Compartments:* ['s1', 's2', 's3', 's4', 's5']
 - *Parameters:* [input.1, input.0, input.0, input.0, input.0]
 - *Exclude From Normalize:* []

4.2 K Matrix

- **km1:**
 - *Label:* km1
 - *Matrix:*
 - * ('s2', 's1'): kinetic.1
 - * ('s5', 's1'): kinetic.2
 - * ('s2', 's2'): kinetic.3
 - * ('s3', 's2'): kinetic.4
 - * ('s4', 's2'): kinetic.5
 - * ('s2', 's3'): kinetic.7
 - * ('s3', 's3'): kinetic.6
 - * ('s2', 's4'): kinetic.8
 - * ('s4', 's4'): kinetic.6
 - * ('s5', 's5'): kinetic.6

4.3 Irf

- **irf1** (gaussian):
 - *Label:* irf1
 - *Type:* gaussian
 - *Center:* irf.center
 - *Width:* irf.width
 - *Normalize:* False
 - *Backsweep:* True
 - *Backsweep Period:* irf.backsweep

4.4 Dataset

- **dataset1:**
 - *Label:* dataset1
 - *Megacomplex:* ['mc1']
 - *Initial Concentration:* input1
 - *Irf:* irf1

4.5 Megacomplex

- **mc1:**
 - *Label:* mc1
 - *K Matrix:* ['km1']

4.6 Spectral Constraints

- **zero:**
 - *Type:* zero
 - *Compartment:* s1
 - *Interval:* [[100, 1000]]
- **zero:**
 - *Type:* zero
 - *Compartment:* s3
 - *Interval:* [[100, 680]]
- **zero:**
 - *Type:* zero
 - *Compartment:* s4
 - *Interval:* [[100, 690]]

4.6.1 Create scheme and optimize it

```
[16]: target_scheme = Scheme(target_model,target_parameters,{'dataset1':_
↪dataset},non_negative_least_squares=True)
target_result = optimize(target_scheme)
```

Iteration	Total nfev	Cost	Cost reduction	Step norm
Optimality				
0	1	1.3863e+08		
1.20e+11				
1	2	4.8613e+07	9.00e+07	2.65e-01
2.49e+10				
2	3	4.6291e+07	2.32e+06	2.78e-02
5.90e+09				
3	4	4.6058e+07	2.33e+05	1.30e-02
2.13e+09				
4	5	4.6049e+07	8.58e+03	2.76e-03
2.53e+08				
5	6	4.6049e+07	5.01e+02	7.49e-04
1.09e+08				
6	7	4.6049e+07	3.21e+01	1.90e-04
2.18e+07				
7	8	4.6049e+07	2.07e+00	5.00e-05
6.48e+06				
8	9	4.6049e+07	1.34e-01	1.26e-05
1.52e+06				
`ftol` termination condition is satisfied.				

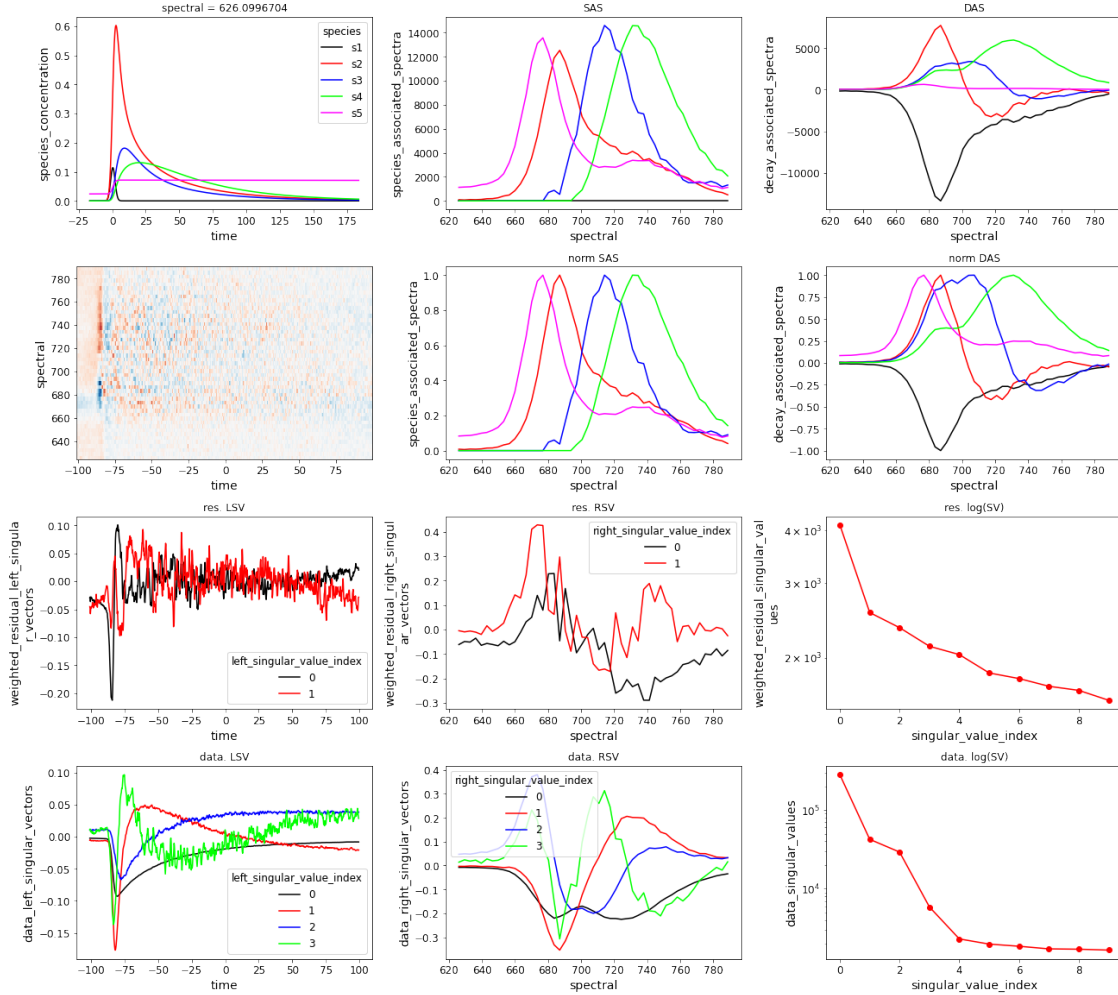
Function evaluations 9, initial cost 1.3863e+08, final cost 4.6049e+07, first-order optimality 1.52e+06.

```
[17]: target_result.data['dataset1']
```

```
[17]: <xarray.Dataset>
Dimensions:                                (clp_label: 5, component: 5,
from_species: 5, left_singular_value_index: 49, right_singular_value_index: 49,
singular_value_index: 49, species: 5, spectral: 49, time: 923, to_species: 5)
Coordinates:
  * time                                (time) float64 -101.0 ... 99.46
  * spectral                            (spectral) float64 626.1 ... 788.7
  * clp_label                           (clp_label) <U2 's1' 's2' ... 's5'
  * species                             (species) <U2 's1' 's2' ... 's5'
    rate                                (component) float64 -2.1 ... -0...
    lifetime                            (component) float64 -0.4762 ...
  * to_species                           (to_species) <U2 's1' ... 's5'
  * from_species                         (from_species) <U2 's1' ... 's5'
Dimensions without coordinates: component, left_singular_value_index,
right_singular_value_index, singular_value_index
Data variables: (12/24)
    data                                (time, spectral) float64 50.75 ...
    data_left_singular_vectors          (time, left_singular_value_index)
float64 ...
    data_singular_values                (singular_value_index) float64 ...
    data_right_singular_vectors         (spectral,
right_singular_value_index) float64 ...
    matrix                              (spectral, time, clp_label)
float64 ...
    clp                                (spectral, clp_label) float64 0...
    ...                                ...
    a_matrix                            (component, species) float64 1...
    k_matrix                            (to_species, from_species) float64
...
    k_matrix_reduced                    (to_species, from_species) float64
...
    irf_center                          float64 -84.26
    irf_width                           float64 1.602
    irf                                 (time) float64 2.23e-24 ... 0.0
Attributes:
    root_mean_square_error:              45.125714328556704
    weighted_root_mean_square_error:     45.125714328556704
```

4.6.2 Result plots

```
[18]: fig = plot_overview(target_result.data['dataset1'],linlog=False)
```



```
[19]: print_md(target_result.optimized_parameters)
```

- None:
 - input:
 - * **1**: Value: 1.0, StdErr: 0.0, Min: -inf, Max: inf, Vary: False, Non-Negative: False, Expr: None
 - * **0**: Value: 0.0, StdErr: 0.0, Min: -inf, Max: inf, Vary: False, Non-Negative: False, Expr: None
 - irf:
 - * **center**: Value: -84.26178625143598, StdErr: 6.008387885244374e-05, Min: -inf, Max: inf, Vary: True, Non-Negative: False, Expr: None
 - * **width**: Value: 1.6016791856249588, StdErr: 6.528613149419845e-05, Min: -inf, Max: inf, Vary: True, Non-Negative: False, Expr: None
 - * **backsweep**: Value: 13800.0, StdErr: 0.0, Min: -inf, Max: inf, Vary: False, Non-Negative: False, Expr: None
 - kinetic:
 - * **1**: Value: 2.0, StdErr: 0.0, Min: -inf, Max: inf, Vary: False, Non-Negative: False,

- Expr*: None
- * **2**: *Value*: 0.1, *StdErr*: 0.0, *Min*: -inf, *Max*: inf, *Vary*: False, *Non-Negative*: False, *Expr*: None
 - * **3**: *Value*: 0.06807153533219637, *StdErr*: 3.366311112870159e-06, *Min*: -inf, *Max*: inf, *Vary*: True, *Non-Negative*: False, *Expr*: None
 - * **4**: *Value*: 0.07244602931988846, *StdErr*: 1.1997720708429068e-05, *Min*: -inf, *Max*: inf, *Vary*: True, *Non-Negative*: False, *Expr*: None
 - * **5**: *Value*: 0.02812014394482191, *StdErr*: 2.8859355608622008e-06, *Min*: -inf, *Max*: inf, *Vary*: True, *Non-Negative*: False, *Expr*: None
 - * **6**: *Value*: 0.0001590654728409451, *StdErr*: 2.173461459677609e-08, *Min*: -inf, *Max*: inf, *Vary*: True, *Non-Negative*: False, *Expr*: None
 - * **7**: *Value*: 0.13547407482819143, *StdErr*: 0.0, *Min*: -inf, *Max*: inf, *Vary*: False, *Non-Negative*: False, *Expr*: \$kinetic.4 * \$scaling.1
 - * **8**: *Value*: 0.03852459720440602, *StdErr*: 0.0, *Min*: -inf, *Max*: inf, *Vary*: False, *Non-Negative*: False, *Expr*: \$kinetic.5 * \$scaling.2
- **scaling**:
- * **1**: *Value*: 1.87, *StdErr*: 0.0, *Min*: -inf, *Max*: inf, *Vary*: False, *Non-Negative*: False, *Expr*: None
 - * **2**: *Value*: 1.37, *StdErr*: 0.0, *Min*: -inf, *Max*: inf, *Vary*: False, *Non-Negative*: False, *Expr*: None

[]: