

# PS1\_pyglotaran

April 21, 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  $i$

Reduced 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, 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]: import matplotlib.pyplot as plt
from pyglotaran_extras.plotting.plot_overview import plot_overview
from pyglotaran_extras.plotting.plot_svd import plot_svd

plt.rcParams["figure.figsize"] = (21, 14)
```

### 2.1.4 Analysis functions

```
[3]: from glotaran.analysis.optimize import optimize
      from glotaran.io import load_dataset, load_model, load_parameters
      from glotaran.project.scheme import Scheme
```

### 2.1.5 Read data

```
[4]: dataset = load_dataset("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 = load_model("models/global-model.yml")
      global_parameters = load_parameters("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, 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:* True
  - *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 scheme 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         (right_singular_value_index,
spectral) 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
```

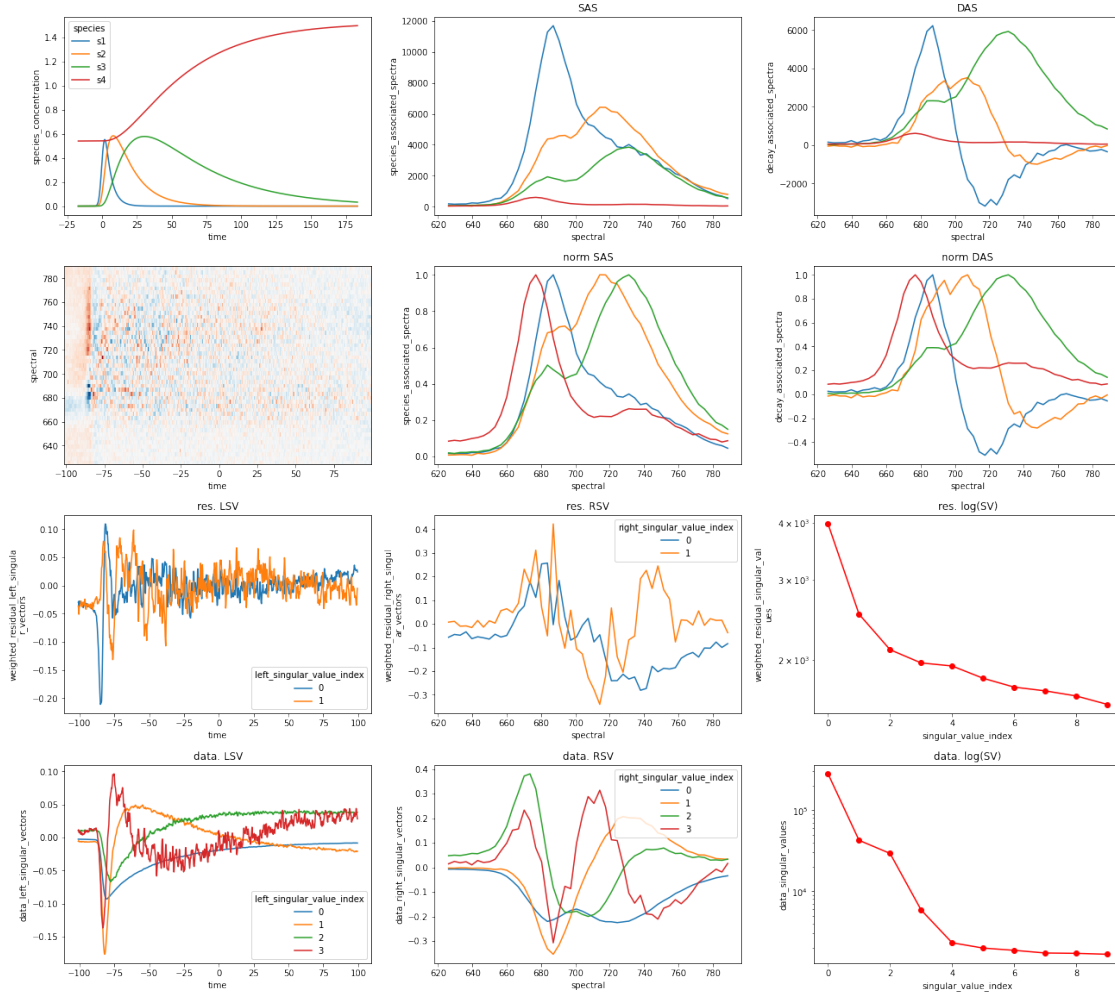
```

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)
```

- input:

Label	Value	StdErr	Min	Max	Vary	Non-Negative	Expr
1	1	0	-inf	inf	False	False	None

<i>Label</i>	<i>Value</i>	<i>StdErr</i>	<i>Min</i>	<i>Max</i>	<i>Vary</i>	<i>Non-Negative</i>	<i>Expr</i>
0	0	0	-inf	inf	False	False	None

- **irf:**

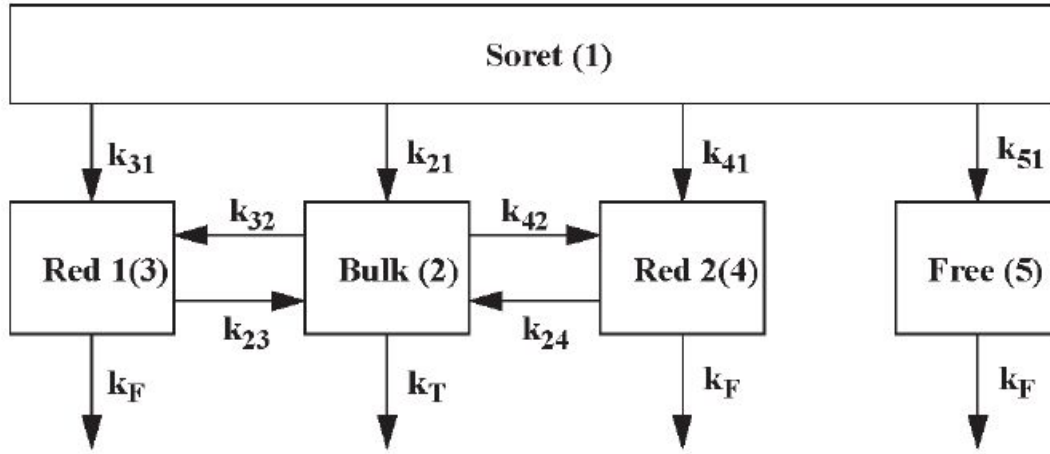
<i>Label</i>	<i>Value</i>	<i>StdErr</i>	<i>Min</i>	<i>Max</i>	<i>Vary</i>	<i>Non-Negative</i>	<i>Expr</i>
center	-83.8533	6.73391e-05	-inf	inf	True	False	None
width	1.60986	6.95281e-05	-inf	inf	True	False	None
backsweep	13200	0	-inf	inf	False	False	None

- **kinetic:**

<i>Label</i>	<i>Value</i>	<i>StdErr</i>	<i>Min</i>	<i>Max</i>	<i>Vary</i>	<i>Non-Negative</i>	<i>Expr</i>
1	0.224983	3.32612e-05	-inf	inf	True	False	None
2	0.0680691	1.19792e-05	-inf	inf	True	False	None
3	0.0212251	1.18649e-06	-inf	inf	True	False	None
4	0.000159693	2.24382e-08	-inf	inf	True	False	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

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

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

$$j(t) = i(t) [1 \ 0 \ 0 \ 0 \ 0]^T$$

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.yml

```
[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 = load_model("models/target-model.yml")
target_parameters = load_parameters("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*: True
  - *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'
```

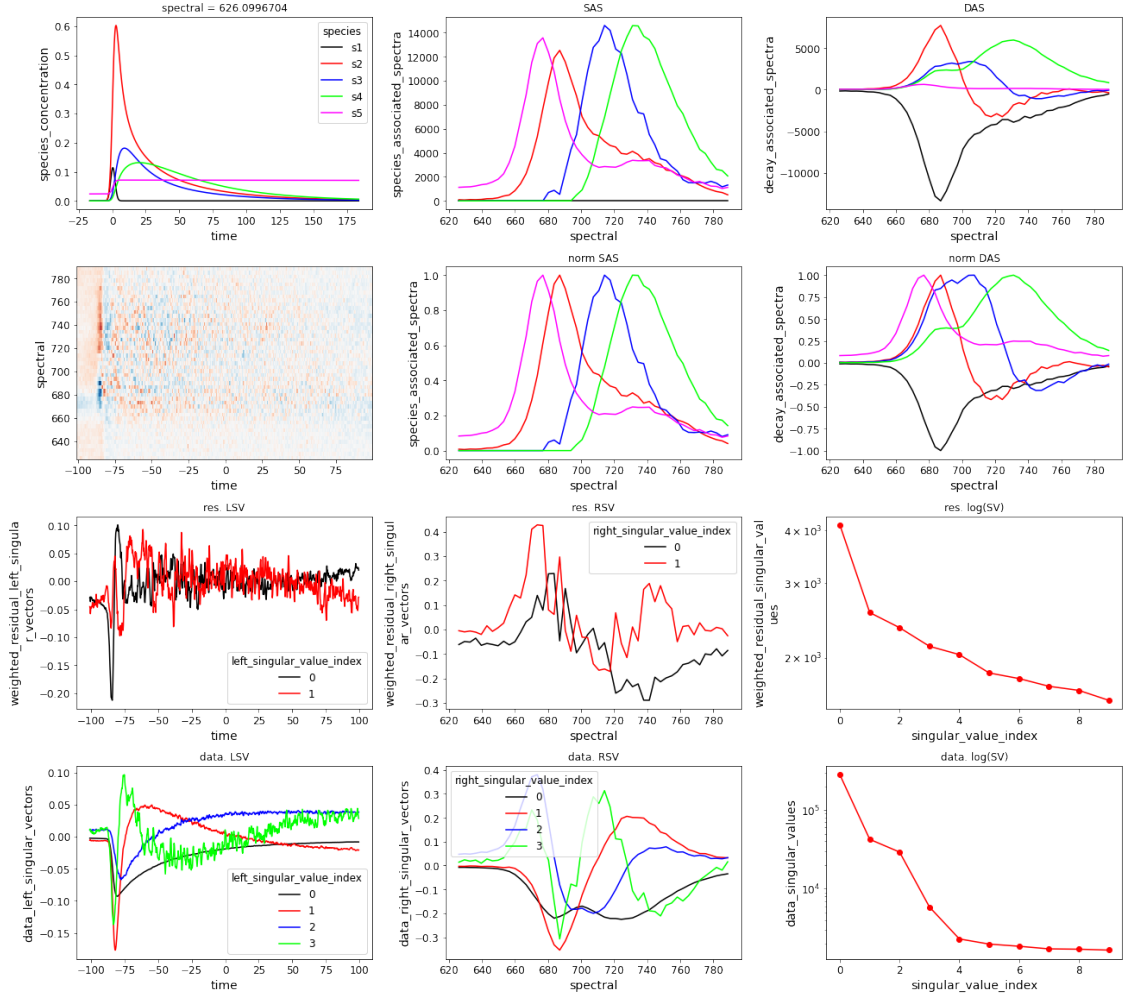
```

    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 (right_singular_value_index,
spectral) 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)
```

- input:

<i>Label</i>	<i>Value</i>	<i>StdErr</i>	<i>Min</i>	<i>Max</i>	<i>Vary</i>	<i>Non-Negative</i>	<i>Expr</i>
1	1	0	-inf	inf	False	False	None
0	0	0	-inf	inf	False	False	None

- irf:

<i>Label</i>	<i>Value</i>	<i>StdErr</i>	<i>Min</i>	<i>Max</i>	<i>Vary</i>	<i>Non-Negative</i>	<i>Expr</i>
center	-84.2618	6.00839e-05	-inf	inf	True	False	None
width	1.60168	6.52861e-05	-inf	inf	True	False	None
backsweep	13800	0	-inf	inf	False	False	None

- **kinetic:**

<i>Label</i>	<i>Value</i>	<i>StdErr</i>	<i>Min</i>	<i>Max</i>	<i>Vary</i>	<i>Non-Negative</i>	<i>Expr</i>
1	2	0	-inf	inf	False	False	None
2	0.1	0	-inf	inf	False	False	None
3	0.0680715	3.36631e-06	-inf	inf	True	False	None
4	0.072446	1.19977e-05	-inf	inf	True	False	None
5	0.0281201	2.88594e-06	-inf	inf	True	False	None
6	0.000159065	2.17346e-08	-inf	inf	True	False	None
7	0.135474	0	-inf	inf	False	False	\$kinetic.4 * \$scaling.1
8	0.0385246	0	-inf	inf	False	False	\$kinetic.5 * \$scaling.2

- **scaling:**

<i>Label</i>	<i>Value</i>	<i>StdErr</i>	<i>Min</i>	<i>Max</i>	<i>Vary</i>	<i>Non-Negative</i>	<i>Expr</i>
1	1.87	0	-inf	inf	False	False	None
2	1.37	0	-inf	inf	False	False	None

[ ]: