Systems biophysics: Global and Target Analysis of Light Harvesting and Photochemical Quenching in vivo

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1 Summary

Systems biophysics can describe the input to a photosynthetic system, and thus contribute a building block to the systems biology of photosynthesis. Systems biophysics mathematically describes the processes during the light reactions of photosynthesis: light harvesting, energy transfer, photochemical quenching (charge separation), nonphotochemical quenching, and state transitions. The aim of *Systems biophysics* is to develop mathematical models that describe the functioning of complex photosynthetic systems. Such models are based upon measured time-resolved absorption and emission spectra, which are two dimensional data sets. Theoretical methods and software have been developed to identify the model, and estimate the biophysical parameters that describe all data. These methods are termed global and target analysis. Key ingredients are compartmental, spectral and thermodynamic models. The methods will be demonstrated with case studies of cyanobacterial light harvesting and photochemical quenching *in vivo*.

Abbreviations: Chl – Chlorophyll; DAS – Decay Associated Spectrum; EAS – Evolution Associated Spectrum; EET – excitation energy transfer; FWHM – Full Width at Half Maximum; IRF – instrument response function; LHC – Light-Harvesting Complex; MA – Magic Angle; $n\lambda$ – number of wavelengths; nt – number of time points; PS – photosystem; RC – reaction center; SAS – Species Associated Spectrum; SVD – singular value decomposition; TCSPT – Time-Correlated Single Photon Timing

2 Introduction

Systems theory methods are widely used in biophysics to describe the functioning of systems with the help of mathematical models. Light absorbed by a light harvesting complex (LHC) provides the input to a photosynthetic system, e.g. the thylakoid membrane. The final outputs of the thylakoid membrane are energized molecules: reduction equivalents (NADPH) and ATP. This chapter will be limited to light harvesting, energy transfer, and photochemical quenching (charge separation) in photosystems (van Grondelle et al. 1994; Croce and van Amerongen 2013; van Amerongen and Croce 2013).

Cyanobacteria are model systems for studies of photosynthesis. They can be mutated which allows to study the properties of subsystems *in vivo*. In particular, the BE and PAL mutant of Synechocystis (Krumova et al. 2010; Ajlani and Vernotte 1998; Tian et al. 2013) can be considered elementary photosynthetic systems. They lack the phycobilisome antenna. The BE mutant only contains photosystem I (PSI) complexes, whereas the PAL mutant contains both PSI and photosystem II (PSII) complexes. Time-resolved spectroscopy has proven to be extremely useful in photosynthesis research in the past decades. Both absorption and emission spectroscopy have been employed, in a variety of wavelength ranges. When the superposition principle applies the system can be considered linear. After excitation by a short laser flash, the (excited state) dynamics of the system can be studied by measuring a time-resolved spectrum. In the language of linear systems theory, the time-resolved spectrum can be considered the response of the system, which is a convolution of the impulse response (the response after a unit impulse of negligible width) and the instrument response function (IRF) (van Stokkum et al. 2004).

This chapter will be limited to measurements of chlorophyll *a* (Chl *a*) emission from PSI or PSII at magic angle. For analysis of polarized-light experiments where anisotropy comes into play, and for transient absorption experiments the reader is referred to (van Stokkum et al. 2004). Time resolved emission spectra can be measured with low excitation intensities, thus avoiding nonlinear effects like annihilation, which can be problematic with transient absorption experiments (Muller et al. 1996).

3 First data: the time-resolved emission spectrum of BE cells

To monitor the fluorescence a synchroscan streak camera in combination with a spectrograph is employed. It can simultaneously record the temporal dynamics and wavelength of fluorescence representable as an image with time and wavelength along the axes (van Stokkum et al. 2008). The time-resolved spectrum of BE cells at room temperature after 400 nm excitation is depicted in Figure 1. The emission evolves both temporally and spectrally: it rises due to the finite rise time of the instrument, the spectral shape evolves due to the equilibration between Chl pigments with different properties, and finally it decays due to photochemical quenching (charge separation, trapping) in the reaction center (RC) (Gobets et al. 2001).



Figure 1. Filled contour plot of the time-resolved emission spectrum of BE cells (containing only PSI) at room temperature after 400 nm excitation. Time step ≈0.8 ps, wavelength step ≈2 nm.

3.1 Usage of the singular value decomposition

The matrix structure of the streak data enables the usage of matrix decomposition techniques, in particular the singular value decomposition (SVD) (Hendler and Shrager 1994; Shrager 1986; Henry and Hofrichter 1992). Formally the data matrix can be decomposed as (Golub and Van Loan 1996)

$$\psi(t,\lambda) = \sum_{l=1}^{m} u_l(t) s_l w_l(\lambda)$$

where u_l and w_l are the left and right singular vectors, s_l the sorted singular values, and *m* is the minimum of the number of rows and columns of the data matrix. The singular vectors are orthogonal, and provide an optimal least squares approximation of the matrix. From the SVD the rank of the data matrix can be estimated, as judged from the singular values and singular vector pairs significantly different from noise. This rank corresponds to the number of spectrally and temporally independent components n_{comp} . When the data matrix has not been corrected for dispersion, this is no longer strictly true, and then the singular values provide only an indication of n_{comp} . Furthermore the SVD of the matrix of residuals is useful to diagnose shortcomings of the model used, or systematic errors in the data.



Figure 2. SVD of the PSI data matrix (a-c) and matrix of residuals (d-i). (a) first three (order squares, circles, triangles) left singular vectors u_l , (b) first three right singular vectors w_l , (c) first ten singular values s_l on a logarithmic scale, Panels (d-f) represent the matrix of residuals from a fit using a simple Gaussian IRF: the first left singular vector $u_{res,1}$, the first right singular vector $w_{res,1}$, and the first ten singular values $s_{res,l}$ on a logarithmic scale. Panels (g-i) represent the matrix of residuals from a fit using a triple Gaussian IRF. Further explanation in text.

Figure 2a-c depicts the SVD of the PSI data, where three singular values and singular vector pairs are significantly different from noise. These first three singular values account for 96.5% of the variance of the data matrix. The left and right singular vectors are both linear combinations of the true concentration profiles and SAS, and are hard to interpret. The first pair (squares) represents a kind of average. The SVD of the residual matrix will be discussed below.

4 Global and target analysis

The aim of global data analysis in general is to obtain a model-based description of the *full* data set in terms of a model containing a small number of precisely estimated parameters. Our main assumption here is that the time and wavelength properties of the system of interest are *separable*, which means that the spectra of species or states are independent of time. Analogously, the dynamics of species or states are assumed to be wavelength independent. When applicable, a wavelength dependence of the IRF

location can be described parametrically (see below). Thus the adjective *global* (Beechem 1989; Beechem et al. 1985) indicates that the kinetic and spectral parameters of the model apply to the *full* data set. The adjective *target* (Arcioni and Zannoni 1984) refers to the targeted model.

A description of the basic ingredient of kinetic models, the exponential decay, will be given first, followed by a description of how to use these ingredients for global and target analysis (see e.g. the reviews by (Holzwarth 1996) and (van Stokkum et al. 2004)) of the full data. The parameter estimation is generally based on nonlinear least squares. Because of the wealth of conditionally linear parameters the variable projection algorithm is crucial (Golub and LeVeque 1979; Mullen and van Stokkum 2009; Nagle 1991; Golub and Pereyra 2003; Golub and Pereyra 1973). For further details on parameter estimation techniques the reader is also referred to the above cited reviews and references cited therein, and to (van Stokkum 2005). Software issues are discussed in (Snellenburg et al. 2012; van Stokkum and Bal 2006; Mullen and van Stokkum 2007).

4.1 Modeling an exponential decay

Here an expression is derived for describing the contribution of an exponentially decaying component to the streak image. The instrument response function (IRF) i(t) can usually adequately be modeled with a Gaussian with parameters μ and Δ for, respectively, location and full width at half maximum (FWHM):

$$i(t) = \frac{1}{\widetilde{\Delta}\sqrt{2\pi}} \exp(-\log(2)(2(t-\mu)/\Delta)^2)$$

where $\tilde{\Delta} = \Delta/(2\sqrt{2\log(2)})$. The adequacy of the Gaussian approximation of the IRF shape is depicted in Figure 3a of (van Stokkum et al. 2008). The convolution (indicated by an *) of this IRF with an exponential decay (with rate *k*) yields an analytical expression which facilitates the estimation of the IRF parameters μ and Δ :

$$c(t,k,\mu,\Delta) = \exp(-kt) * i(t) = \frac{1}{2} \exp(-kt) \exp(k(\mu + \frac{k\widetilde{\Delta}^2}{2})) \{1 + erf(\frac{t - (\mu + k\widetilde{\Delta}^2))}{\sqrt{2\widetilde{\Delta}}}\}$$

The periodicity of the synchroscan results in detection of the fluorescence that remains after multiples of half the synchroscan period T (typically T \approx 13 ns). Therefore, if lifetimes longer than \approx 1 ns occur in a sample, the above expression should be extended with a summation over the signal contributions that result from forward and backward sweeps:

$$c(t,k,T) = \sum_{n=0}^{\infty} e^{-kTn} \left\{ e^{-k(t-\mu+T)} + e^{-k(T/2-t-\mu)} \right\} = \left\{ e^{-k(t-\mu+T)} + e^{-k(T/2-t-\mu)} \right\} / (1-e^{-kT})$$

Note that it is assumed here that time zero of the time base corresponds to the zero crossing of the sweep, and that the convolution with the IRF is no longer necessary at times longer than T/2. Adding the previous expressions provides the full model function for an exponential decay recorded with a synchroscan streak camera and will henceforth be denoted by c'(k):

$$c^{I}(k) \equiv c(t,k,\mu,\Delta,T) = c(t,k,\mu,\Delta) + c(t,k,T)$$

Because fluorescence samples are relatively dilute, elastic scattering or Raman scattering of the excitation light by water (or of other solvents) can complicate the measurement, if they occur within the analyzed wavelength interval. Such contributions can be modeled with an extra component with a time course identical to the IRF i(t). Usually it is possible to restrict the contribution of scattering to a limited wavelength region.

If the streak image has not been corrected for the instrumental curvature the wavelength dependence of the IRF location μ can be modeled with a polynomial (usually a parabola is adequate). Sometimes the IRF shape is better described by a superposition of two or even three Gaussians (with common location μ), leading to a superposition description of the exponential decay (van Stokkum, 2005). We will return to this below.

The above IRF description for the synchroscan streak camera is fairly complicated. With time-resolved absorption spectroscopy the Gaussian approximation of the IRF shape describes the pump-probe overlap rather well. With time-correlated single photon timing (TCSPT) the IRF can be measured via scatter (at the excitation wavelength) or with a fast decaying reference compound (at the detection wavelength). In practice, the TCSPT IRF can be well approximated by a superposition of three or four Gaussians (with different locations μ).

4.2 The superposition principle

The basis of global analysis is the superposition principle, which states that the measured data $\psi(t,\lambda)$ result from a superposition of the spectral properties $\varepsilon_l(\lambda)$ of the components present in the system of interest weighted by their concentration $c_l(t)$.

$$\psi(t,\lambda) = \sum_{l=1}^{n_{comp}} c_l(t) \varepsilon_l(\lambda)$$

The $c_l(t)$ of all n_{comp} components are described by a compartmental model, that consists of first-order differential equations, with as solution sums of exponential decays. We will consider three types of compartmental models: (1) a model with components decaying monoexponentially in parallel, which yields Decay Associated Spectra (DAS), (2) a sequential model with increasing lifetimes, also called an unbranched unidirectional model, giving rise to Evolution Associated Spectra (EAS), and (3) a full compartmental scheme which may include possible branchings and equilibria, yielding Species Associated Spectra (SAS). The latter is most often referred to as target analysis, where the target is the proposed kinetic scheme, including possible spectral assumptions.

4.3 Parallelly decaying components: DAS

With parallelly decaying components the model reads

$$\psi(t,\lambda) = \sum_{l=1}^{n_{comp}} c^{l}(k_{l}) DAS_{l}(\lambda)$$

The DAS thus represent the estimated amplitudes of the above defined exponential decays $c^{I}(k_{I})$.



Figure 3. Results from global analysis of BE mutant (PS I) data depicted in Figure 1. Note that in a–c and e the time axis is linear from –70 to +70 ps relative to the maximum of the IRF, and logarithmic thereafter. Insets in a, b show residuals. (a) Data (in grey) and fit (in black) of 685 nm emission trace showing multiexponential decay. Contributions of the three exponential decays with different lifetimes (shown in c) multiplied by their amplitudes at 685 or 710 nm (shown in d) are indicated by line type in panels a and b. (b) Data (in grey) and fit (in black) of 710 nm emission trace showing multiexponential rise (visible as a contribution of $c^{I}(k_{1})$ with negative amplitude) and decay. (c) Exponential decays $c^{I}(k_{1})$. Estimated lifetimes (in panels c and e): 5.8 ps (solid), 26.4 ps (dashed), 0.59 ns (dotted). Note that in panels c and e the dotted concentration profile was divided by 25. (d) Estimated Decay Associated Spectra (DAS). Note that in panels d and f the dotted spectrum was multiplied by 25 to increase its visibility. (e) Evolutionary concentration profiles c_{I}^{II} (assuming a sequential kinetic scheme with increasing lifetimes). (f) Estimated Evolution Associated Spectra (EAS).

Three components are sufficient to describe the data. The adequacy of the fit is judged from the SVD of the first left and right singular vectors of the matrix of residuals. With a simple Gaussian IRF shape the residuals (shown in Figure 2d-f) show some structure, especially at times before the maximum of the IRF, and at all wavelengths where there is signal. This indicates that the IRF shape is more complicated. Therefore we applied a triple Gaussian IRF shape with FWHM 9 ps (61% of the area), 15 ps (29%) and 204 ps (10%). We attribute these small fractions of wider IRF to shortcomings of the synchroscan. The trends are no longer present in Figure 2g,h. The rms error of the fit decreased from 1.154 to 1.136, which is just over 1% of the peak in Figure 1.

The estimated lifetimes (reciprocals of estimated rate constants k_l) are (5.8±0.2) ps, (26.4±0.2) ps, and (0.59±0.02) ns.

The estimated DAS from the PSI data are shown in Figure 3d. Several observations can be made: the 5.8 ps DAS (solid) is nearly conservative, i.e., the positive and negative areas are almost equal. It represents decay of more blue and rise of more red emission, and can be interpreted as energy transfer from Bulk to Red Chl *a*, i.e. Chl *a* that absorbs at wavelengths longer than the primary electron donor P700. Thus *in vivo* the excited state equilibration time constant of PSI is 5.8 ps (Krumova et al. 2010; Chukhutsina et al. 2013). The 26.4 ps DAS represents the trapping spectrum. The long lived (0.59 ns) DAS (dotted) was multiplied by 25 to increase its visibility. It has a maximum near 670 nm and resembles the shape of a Chl *a* fluorescence spectrum. Most probably it is not connected to PSI. It could be the single Chl *a* present in cytochrome b6f (Peterman et al. 1998). Henceforth it will be referred to as "free" Chl *a*. Note that in a steady-state spectrum its contribution will be as large as that of PSI, because of its long lifetime. Clearly, the first two DAS do not represent pure species, and they are interpreted as linear combinations (with positive and negative contributions) of true species spectra.

Note that the ultrafast (<<1 ps) rise due to the relaxation from the initially excited Soret state (higher excited state, of which the \approx 460nm emission is outside the detection range) to the Q_y emission (lowest excited state) cannot be resolved in this experiment due to the limited time resolution (the main FWHM Δ of the IRF was 9 ps).

4.4 Sequentially decaying components: EAS

The sequential model reads

$$\psi(t,\lambda) = \sum_{l=1}^{n_{comp}} c_l^{II} EAS_l(\lambda)$$

where each concentration is a linear combination of the exponential decays (Nagle et

al. 1982),
$$c_l^{II} = \sum_{j=1}^l b_{jl} c^I(k_l)$$
, and the amplitudes b_{jl} are given by $b_{11} = 1$ and for $j \le l$:

$$b_{jl} = \prod_{m=1}^{l-1} k_m / \prod_{n=1, n \neq j}^{l} (k_n - k_j)$$

Examples of c_l^{II} are depicted in Figure 3e, whereas the estimated EAS are shown in Figure 3f. With increasing lifetimes, and thus decreasing rates k_l , the first EAS (equal to the sum of DAS) corresponds to the spectrum at time zero with an ideal infinitely small IRF, $i(t) = \delta(t)$. The first EAS (solid line in Figure 3f) represents the sum of the spectra of all excitations that have arrived from the Soret region, and is dominated by Bulk Chl a. The second EAS, which is formed with a time constant of 5.8 ps and decays with a time constant of 26.4 ps, contains more Red Chl a emission, and represents a Boltzmann equilibrium of all PSI Chl a pigments. In this interpretation we neglect the small contribution of the long lived (0.59 ns) "free Chl a". When we collate the concentration profiles of the parallel and sequential model in matrices C¹ and C^{II} , with *nt* rows and n_{comp} columns, respectively, these obey the matrix relation $C^{II} = C^{I}B$, with the elements of the $n_{comp} \times n_{comp}$ upper triangular B matrix defined above. Likewise, we form $n\lambda \times n_{comp}$ matrices DAS and EAS that contain the DAS₁ and EAS_1 of the components. In matrix form the superposition model for the as $\Psi = C^{I} \cdot DAS^{T}$ (parallel) $nt \times n\lambda$ data matrix Ψ is expressed or $\Psi = C^{II} \cdot EAS^{T}$ (sequential). In global analysis (without any spectral constraints, see below), the parallel and sequential model result in *exactly* the same residuals and exactly the same estimated lifetimes. The estimated concentration profiles obey and therefore the estimated DAS and EAS are related as $C^{II} = C^{I}B$ $C^{II} \cdot EAS^T = C^I B \cdot EAS^T = C^I \cdot DAS^T$ which leads to $B \cdot EAS^T = DAS^T$ or when

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transposed $EAS \cdot B^T = DAS$ or $EAS = DAS \cdot B^{-T}$. The lower triangular matrix B^T can be inverted yielding $b_{1l}^{-1} = 1 = b_{l1}^{-T}$ and

$$b_{jl}^{-1} = \prod_{n=1}^{j-1} \frac{(k_n - k_l)}{k_n} = b_{lj}^{-T}$$

For the three components of the above global analysis we can compute (inserting $k_1 = 0.173$, $k_2 = 0.038$, $k_3 = 0.0016$)

$$B = \begin{bmatrix} 1 & -k_1/(k_1 - k_2) & k_1/(k_1 - k_2) \cdot k_2/(k_1 - k_3) \\ 0 & k_1/(k_1 - k_2) & k_1/(k_1 - k_2) \cdot k_2/(k_3 - k_2) \\ 0 & 0 & k_1/(k_1 - k_3) \cdot k_2/(k_2 - k_3) \end{bmatrix} = \begin{bmatrix} 1 & -1.28 & 0.28 \\ 0 & 1.28 & -1.34 \\ 0 & 0 & 1.06 \end{bmatrix}$$
$$B^{-T} = \begin{bmatrix} 1 & 0 & 0 \\ 1 & (k_1 - k_2)/k_1 & 0 \\ 1 & (k_1 - k_3)/k_1 & (k_1 - k_3)/k_1 \cdot (k_2 - k_3)/k_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 0.78 & 0 \\ 1 & 0.99 & 0.95 \end{bmatrix}$$

This leads to the following relations for the DAS_1 and EAS_1 of the components: $EAS_1 = DAS_1 + DAS_2 + DAS_3$, $EAS_2 = 0.78DAS_2 + 0.99DAS_3$, $EAS_3 = 0.95DAS_3$, $DAS_1 = EAS_1 - 1.28EAS_2 + 0.28EAS_3$, $DAS_2 = 1.28EAS_2 - 1.34EAS_3$

4.5 Target model: SAS

When neither of these two simple models is applicable, a full kinetic scheme may be appropriate. This is called a compartmental model. The problem with most kinetic schemes is that while the kinetics are described by more than n_{comp} microscopic rate constants (due to branchings or equilibria), the data only allows for the estimation of n_{comp} decay rates (or lifetimes). Then additional information is required to estimate the microscopic rates (Nagle 1991), which can be spectral constraints (zero contribution of SAS at certain wavelengths) or spectral relations. This is detailed in (van Stokkum et al. 2004).

Now the model reads

$$\psi(t,\lambda) = \sum_{l=1}^{n_{comp}} c_l^{III} SAS_l(\lambda)$$

Disregarding the small contribution of the long lived (0.59 ns) "free Chl a" we can describe the PSI dynamics by the following two coupled differential equations (omitting the superscript III):

$$\frac{d}{dt} \begin{bmatrix} c_1(t) \\ c_2(t) \end{bmatrix} = \begin{bmatrix} -(k_{01} - k_{21}) & k_{12} \\ k_{21} & -(k_{02} - k_{12}) \end{bmatrix} \begin{bmatrix} c_1(t) \\ c_2(t) \end{bmatrix} + \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} i(t)$$

where the indices 1 and 2 refer to Bulk and Red Chl a, respectively. It is known that a PSI monomer contains about six Red Chl a and 90 Bulk Chl a, thus the inputs to the system are $x_1 = 90/96$ and $x_2 = 6/96$. There are four microscopic rate constants, k_{21} and k_{12} representing the equilibration, k_{01} representing the photochemical quenching (trapping) and k_{02} representing the natural lifetime (k_f) of a Chl a. Typically $k_f \approx 1/(2ns)$, but this means that still three microscopic rate constants need to be estimated, whereas only two lifetimes and two SAS_1 can be estimated. Thus spectral constraints imposed on the two SAS_i are needed. When modeling emission data, a natural spectral constraint is nonnegativity of the SAS₁ (Mullen and van Stokkum 2009). However, this constraint is still not enough. A strong spectral constraint is $SAS_2 = 0$ in an appropriate wavelength range (typically below 685 nm for Red Chl *a*). This strong constraint is enough to estimate the equilibrium between the two states (Gobets et al. 2001; Muller et al. 1992). A mild spectral constraint is to assume that the area under the SAS_i is equal for states of equal oscillator strength like Bulk and Red Chl a (Snellenburg et al. 2013). This mild constraint is also enough to estimate the equilibrium between the two states, and is especially useful when a strong zero constraint cannot be applied. Formally, when the ratio between the oscillator strengths of two Chl species with SAS_i and SAS_j is assumed to be α , then a penalty can be imposed which is added to the least squares criterion of the fit:

penalty_{ij} = weight_{ij}
$$\int_{\lambda \min}^{\lambda \max} |SAS_i(\lambda) - \alpha SAS_j(\lambda)| d\lambda$$

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Without a priori knowledge, we assume that all Chl species possess the same oscillator strength, i.e. $\alpha = 1$. Here weight_{ij} is used to tune the importance of this area constraint in the least squares fitting process.

4.5.1 Solving the compartmental model

The general compartmental model with p compartments can be abbreviated as $\frac{d}{dt}c(t) = Kc(t) + j(t) \text{ with } c(t) = \begin{bmatrix} c_1(t) & \dots & c_p(t) \end{bmatrix}^T, \ j(t) = i(t)x \text{ and } x = \begin{bmatrix} x_1 & \dots & x_p \end{bmatrix}^T$ is a vector that describes the amount of excitation of each compartment. In the above example the concentration vector is $c(t) = \begin{bmatrix} c_1(t) & c_2(t) \end{bmatrix}^T$ and the input vector is $j(t) = i(t) \begin{bmatrix} x_1 & x_2 \end{bmatrix}^T$. The transfer matrix K contains off-diagonal elements k_{ij} , representing the microscopic rate constant from compartment j to compartment i. k_{0j} indicates decay from compartment i to the outside of the system described. The diagonal elements contain the total decay rates of each compartment. To solve a compartmental model we employ the eigendecomposition of K given by $U\Lambda U^{-1}$ with $\Lambda = diag(-\kappa_1, -\kappa_2, ..., -\kappa_p)$ the diagonal matrix containing the eigenvalues $-\kappa_l$ of K, and U the matrix with as columns the eigenvectors (check: $KU = U\Lambda U^{-1}U = U\Lambda$, thus each eigenvector is scaled by its eigenvalue). Thus we have $e^{Kt} = Ue^{\Lambda t}U^{-1}$ (since $\exp(Kt) = \sum_{n=0}^{\infty} (Kt)^n / n!$ and $K^n = U\Lambda U^{-1} \dots U\Lambda U^{-1} = U\Lambda^n U^{-1}$ thus $\sum_{n=0}^{\infty} (Kt)^n / n! = U(\sum_{n=0}^{\infty} (\Lambda t)^n / n!) U^{-1} = U \exp(\Lambda t) U^{-1}) \text{ and }$ $c^{III}(t) = e^{\kappa_{t}} * j(t) = U diag(U^{-1}x) \begin{bmatrix} e^{-\kappa_{1}t} * i(t) \\ \dots \\ e^{-\kappa_{p}t} * i(t) \end{bmatrix} = U diag(U^{-1}x) \begin{bmatrix} c^{I}(\kappa_{1}) \\ \dots \\ c^{I}(\kappa_{p}) \end{bmatrix}$

Thus the solution of the general compartmental model is a linear combination of the $c^{I}(\kappa_{i})$. The concentration of each compartment is a linear combination of at most p exponential decays (convolved with the IRF) with coefficients that depend upon the microscopic rate constants that describe the transitions between all the compartments.

4.5.2 Detailed balance

When two compartments *i* and *j* are at equilibrium, we have

$$k_{ij}c_j(t) = k_{ji}c_i(t)$$

and we can compute the Gibbs free energy difference between compartments i and j as

$$\Delta G_{ii} = k_B T \ln(k_{ii} / k_{ii})$$

For a closed system at equilibrium the sum of the Gibbs free energy differences of a cycle equals zero. Thus e.g. with three compartments the product of the clockwise rate constants $1 \rightarrow 2 \rightarrow 3 \rightarrow 1$ is equal to the product of the counterclockwise rate constants $1 \leftarrow 2 \leftarrow 3 \leftarrow 1$:

$$k_{21}k_{32}k_{13} = k_{12}k_{31}k_{23}$$

This is called a detailed balance condition. It must be fulfilled, and reduces the number of unknown microscopic rate constants (Nagle 1991).

4.5.3 Target analysis of the first data

The kinetic scheme used for the target analysis of the BE PSI data is depicted in Figure 4a. As explained above three microscopic rates, the forward and backward rate constants between the Bulk (B) and Red (R) Chl *a* compartment, and the rate of photochemical quenching from B were estimated with the help of the mild spectral constraint that the area under the SAS_1 is equal for states B and R. The shapes of the estimated SAS shown in Figure 4c are realistic. Very small amplitudes of SAS_R are present below 690 nm.

The free energy difference between two compartments depends upon the energy difference between the two emission maxima and upon the relative number of pigments of the compartments ($k_BT \ln(N_1/N_2)$). We estimate the energetic contribution from the difference between the average emission wavelength of the B and R SAS, 703.6 and 722.7 nm for B* and R*, respectively. This corresponds to 47 meV. The observed free energy difference is

 $\Delta G_{12} = k_B T \ln(k_{21} / k_{12}) = k_B T \ln(61/126) = -18 \text{ meV}$. The entropic free energy difference thus equals 47+18=65 meV, which is equal to $k_B T \ln(N_1 / N_2)$, with N_1 and N_2 the number of pigments in the B and R compartments. Using $k_B T = 25meV$ we thus find $N_1 / N_2 = 13.1$ which is consistent with N1 and N2 comprising ≈ 92 and ≈ 7 pigments, respectively. This is very close to the above assumption that the PSI monomer contains about six Red Chl *a* and 90 Bulk Chl *a*.



Figure 4. (a) Kinetic scheme used for the target analysis of the BE PSI data depicted in Figure 1. After excitation in the Soret band three compartments are populated: Bulk Chl *a* (B), Red Chl *a* (R) and a small fraction of "free Chl *a*" (F). The first two compartments equilibrate, and excitations are trapped from B. Rates in 1/ns. (b) concentration profiles c_l^{III} , note that the time axis is linear from -70 to +70 ps relative to the maximum of the IRF, and logarithmic thereafter. (c) Species Associated Spectra (SAS).

Key in a, b and c: Bulk Chl *a* (solid), Red Chl *a* (dashed), "free Chl *a*" (dotted). In addition, panel b depicts the rise of the charge separated state (radical pair) RP1 (reaching 0.96) in dot-dash.

The eigenvalues of the K matrix correspond to lifetimes of 4.7 and 28 ps, and 0.68 ns. These lifetimes differ slightly from those estimated in the global analysis (5.6 and 26.4 ps, and a very small contribution of 0.59 ns). The difference is attributed to the limited signal to noise ratio of the data. Apparently, the relative precision of the estimated lifetimes is typically about 10%. The quality of the fit (rms errors and structure of the residual matrix) of the global and of the target analysis are practically

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the same.

Finally, the amplitude matrix A is presented in Table 1, which expresses the relation between the concentration profiles c_l^{III} and the exponential decays $c^I(\kappa_l)$,

$$C^{III} = A \cdot C^I \, .$$

The interpretation of the amplitude matrix is as follows (neglecting the IRF):

$$c_B = 0.392 \exp(-t/4.7) + 0.508 \exp(-t/28)$$

$$c_R = -0.282 \exp(-t/4.7) + 0.342 \exp(-t/28)$$

$$c_{RP1} = -0.110 \exp(-t / 4.7) - 0.838 \exp(-t / 28) + 0.948$$

The total PSI excitation is 0.96, and the RP1 yield is 0.948/0.96=99%. No attempt is made to resolve the equilibration within the Bulk compartment to and from the RC. Several models are discussed in (van Stokkum et al. 2013; Muller et al. 2010).

Compartment	Red	Bulk	RP1	F
Excitation	0.06	0.90	0	0.024
4.7 ps	-0.282	0.392	-0.110	
28 ps	0.342	0.508	-0.838	
long lived			0.948	
0.68 ns				0.024

Table 1. Amplitude matrix for the different lifetimes of the BE PSI data

5 Second data: the time-resolved emission spectrum of PAL cells

The time-resolved spectrum of PAL cells at room temperature after 400 nm excitation is depicted in Figure 5. The PSIIRC is in the open state (top), or in the closed state, due to the presence of DCMU (bottom). Note that PAL cells contain both PSI and PSII, and thus a target model must be a superposition of target models for both PSI and PSII.

Again, the emission evolves both temporally and spectrally: it rises due to the finite rise time of the instrument, the spectral shape evolves due to equilibration between Chl pigments with different properties, and finally it decays due to photochemical quenching (charge separation, trapping) in the reaction centers (PSIRC and PSIIRC).



Figure 5. Filled contour plot of the time-resolved emission spectrum of PAL cells (that contain both PSI and PSII) at room temperature after 400 nm excitation. Time step \approx 0.8 ps, wavelength step \approx 2 nm. Top: PSIIRC in the open state, bottom PSIIRC in the closed state, due to the presence of DCMU.



Figure 6. Results from global analysis of PAL mutant data with PSIIRC in the open state depicted in Figure 5 top. Note that in a–c and e the time axis is linear from –70 to +70 ps relative to the maximum of the IRF, and logarithmic thereafter. Insets in a, b show residuals. (a) Data (in grey) and fit (in black) of 687 nm emission trace showing multiexponential decay. Contributions of the four exponential decays with different lifetimes (shown in c) multiplied by their amplitudes at 687 or 710 nm (shown in d) are indicated by line type in panels a and b. (b) Data (in grey) and fit (in black) of 710 nm emission trace showing multiexponential rise (visible as a contribution of $c^{T}(k_{1})$ with negative amplitude) and decay. (c) Exponential decays $c^{T}(k_{1})$. Estimated lifetimes (in panels c and e): 4.6 ps (solid), 16.5 ps (dashed), 45.8 ps (dotted) and 382 ps (dot-dashed). (d) Estimated Decay Associated Spectra (DAS). (e) Evolutionary concentration profiles c_{1}^{II} (assuming a sequential kinetic scheme with increasing lifetimes). (f) Estimated Evolution Associated Spectra (EAS).

Four components are sufficient to describe the data. The estimated lifetimes (reciprocals of estimated rate constants k_i) are (4.6±0.2), (16.5±0.3), (45.8±0.5), and (382±3) ps. Judging the DAS is less straightforward than with BE. The fastest lifetime (4.6 ps) again corresponds to equilibration in PSI, where the rise of the red Chl *a* is

evidenced by the negative amplitudes above 690 nm. The other three DAS are all positive, and demonstrate photochemical quenching on three time scales. Apparently both PSI and PSII show quenching on time scales of 16 and 46 ps. The final lifetime of 382 ps can safely be assigned to PSII. The EAS confirm this rough picture. However, target analysis is mandatory to disentangle the contributions of PSI and PSII. From the SVD of the data (not shown) it is clear that three temporally and spectrally different components are present. The fourth pair of left and right singular vectors represents noise. Thus it can be concluded that in these data the PSII dynamics are describable with a single spectrum decaying multiexponentially.

5.2 Simultaneous target analysis of both PAL data sets

Although a target analysis of each individual PAL data set is feasible, it is more instructive to present here a simultaneous analysis, in order to demonstrate that both PAL data sets can be described by a single model that differs only in the parameters describing the open or closed condition of the PSIIRC.



Figure 7. (a) Kinetic scheme used for the target analysis of the PAL data depicted in Figure 5. After excitation in the Soret band three compartments are populated: Bulk Chl *a* (B1), Red Chl *a* (R1) of PSI, and Bulk Chl *a* (B2) of PSII. B2 equilibrates with a non-radiative radical pair (RP2) compartment, thus effectively describing the biexponential decay of the PSII emission. Excitations are trapped from B1 or B2. Rates in 1/ns. (b) concentration profiles C_l^{III} , note that the time axis is linear from -70 to +70 ps relative to the maximum of the IRF, and logarithmic thereafter. (c) Species Associated Spectra (SAS). Key in a, b and c: PSI in grey with B1 (solid), R1 (dashed), PSII in black with (in the closed state) B2 (solid), RP2 (dashed), and (in the open state) B2 (dot-dashed), RP2 (dotted).

Now there are three K matrices. The eigenvalues of the PSI K matrix correspond to lifetimes of 3.7 and 26.6 ps. These lifetimes are slightly shorter than those estimated in the BE target analysis. The estimated PSI SAS of the BE and PAL mutant are practically identical, apart from a small shift due to wavelength calibration uncertainty. Hence the PSI results of the PAL mutant will not be discussed further. The relative contributions of PSI and PSII can be estimated. Assuming that PSI is in trimeric form, and PSII in dimeric form, they contain $3 \times 96 = 288$ and $2 \times 35 = 70$ Chl *a*, respectively. We then estimate a stoichiometry of 1.74 or 1.83 PSII dimer per PSI trimer for the samples with PSIIRC in the open or closed condition, respectively. When expressed as the ratio of PSIRC to PSIIRC this corresponds to 0.86 or 0.82,

which is in agreement with the values reported by (Stadnichuk et al. 2009) of 0.7 ± 0.2 and 0.76 (Krumova et al. 2010; Tian et al. 2013).

Finally, the amplitude matrices A describing the PSII dynamics with PSIIRC in the open or closed condition are presented in Table 2 and Table 3, respectively. The interpretation of the amplitude matrix is as follows (neglecting the IRF):

$$c_{B2,open} = 0.226 \exp(-t/55) + 0.071 \exp(-t/417)$$
,

 $c_{B2,closed} = 0.108 \exp(-t/123) + 0.200 \exp(-t/877)$. Effectively, in PSII cores two lifetimes are sufficient to describe photochemical quenching *in vivo* in the PAL mutant. *In vitro*, two additional short lifetimes (<10 ps) with small contributions can be resolved from time resolved emission spectra of PSII cores of *Thermosynechococcus elongatus* with open PSIIRC (Miloslavina et al. 2006; van der Weij–de Wit et al. 2011). The *in vitro* PSII core emission from (van der Weij–de Wit et al. 2011) can be well fitted with the simple two compartment scheme of bulk and radical pair resulting in $c_{B2,open}^{invitro} = 0.834 \exp(-t/42) + 0.156 \exp(-t/173)$. The

photochemical quenching parameters of PSII are collated in Table 4.

Thus the main differences with open RC are that *in vivo* photochemical quenching is a bit slower than *in vitro*, and that the free energy difference between B2 and RP2 is smaller. The transmembrane potential could be the cause thereof. *In vivo* with closed RC, the negative charge on Q_A^- lifts the RP2 to a higher free energy than that of B2.

 Table 2. Amplitude matrix for the different lifetimes of the PAL data with open PSIIRC

 describing PSII

Compartment	B2	RP2
Excitation	0.297	0
55 ps	0.226	-0.262
417 ps	0.071	0.262

 Table 3. Amplitude matrix for the different lifetimes of the PAL data with closed PSIIRC

 describing PSII

Compartment	B2	RP2
Excitation	0.308	0
123 ps	0.108	-0.137
877 ps	0.200	0.137

sample	$k_{B2 \to RP2}$	$k_{RP2 \rightarrow B2}$	$k_{RP2 \rightarrow}$	$\Delta G_{B2,RP2}$
In vitro, open	20	2.1	6.5	57
In vivo, open	14	3.3	2.9	37
In vivo, closed	3.1	3.6	2.1	-4

Table 4. Estimated photochemical quenching parameters of PSII, rates in 1/ns, $\Delta G_{\scriptscriptstyle B2,RP2}$ in meV

Since it is not possible to resolve the small contributions of the faster equilibration lifetimes (<10 ps) no attempt is made to resolve the equilibration within the B2 compartment between the PSIIRC and the CP43 and CP47 antenna complexes, which is still disputed (Raszewski and Renger 2008; Barter et al. 2001; van der Weij–de Wit et al. 2011; Miloslavina et al. 2006; Tian et al. 2013).

6 Conclusions

Systems biophysics can resolve *in vivo* processes. Combined with knowledge from the literature, simultaneous analysis of different experiments can describe the functioning of complex photosynthetic systems. The high time resolution experiments of *in vitro* systems, preferably with selective excitation of e.g. the reaction center (Holzwarth et al. 2006; Muller et al. 2010; Groot et al. 2005) can reveal details of ultrafast processes that are obscured by the relatively slow dynamics of energy transfer to the reaction center. Cyanobacteria are promising model systems for a full description of the intact thylakoid membrane with the help of target analysis.

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