



Photosynthesis and Phototrophic Growth: Modelling Life on Earth and elsewhere

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Fakulta informatiky MU

pondělí 14. 5. 2012, 14:00





Part I

News and Events on May 14, 2012

Part II

Cyanobacteria: understanding phototrophic growth

Dynamics of large-scale networks

Applications to biotechnology





May 14, 2012, 0am





May 14, 2012, 0am



Water, water, everywhere, ...

Bacterial abundance in in stratified oligotrophic waters can be high ($> 10^5$ cells ml^{-1})



May 14, 2012, 0am



Water, water, everywhere, ...

Bacterial abundance in stratified oligotrophic waters can be high ($> 10^5$ cells ml^{-1})

But no primary productivity ...



May 14, 2012, a few hours later ...





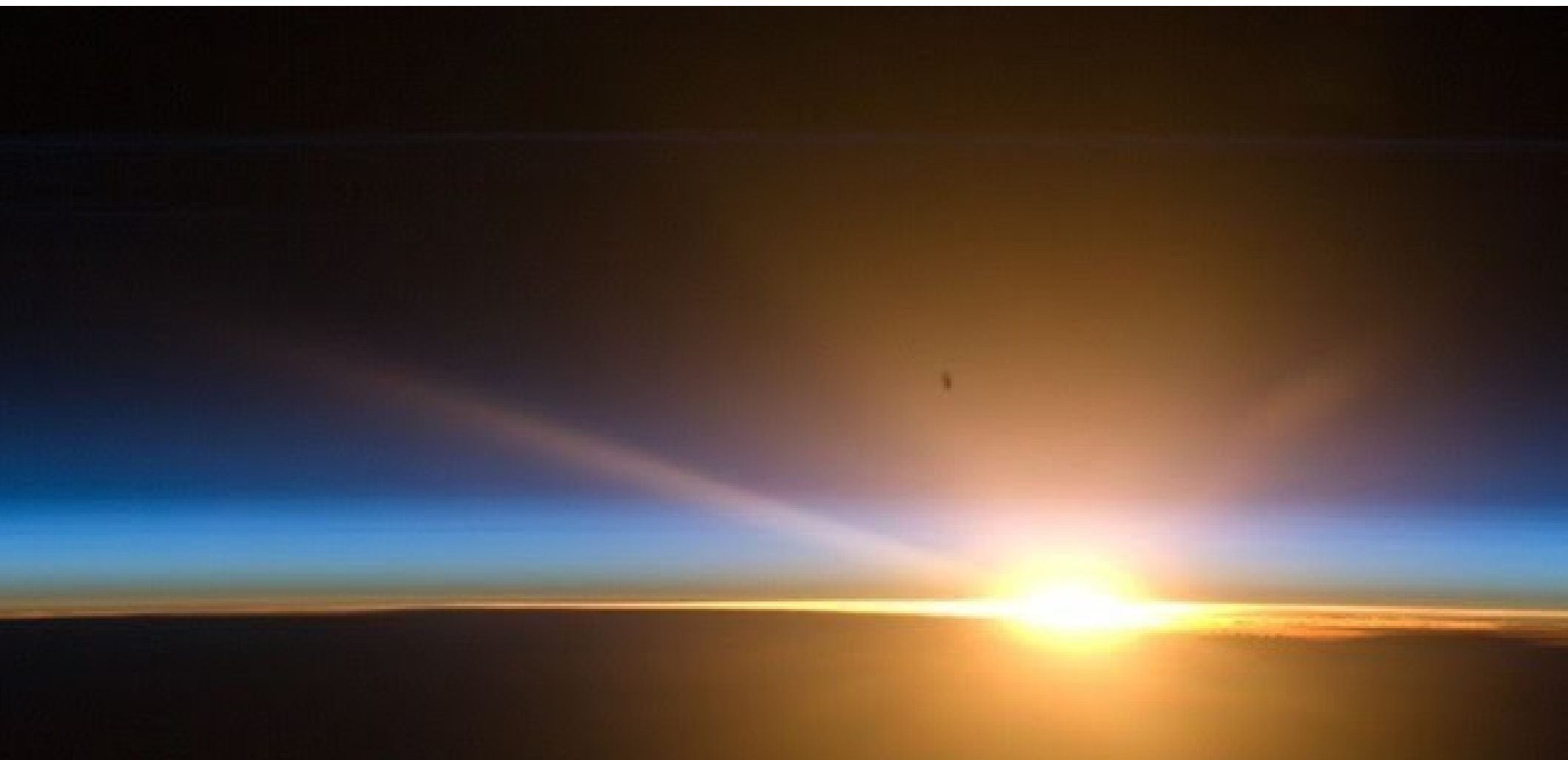
outline

May 14, 2012, a few hours later ...





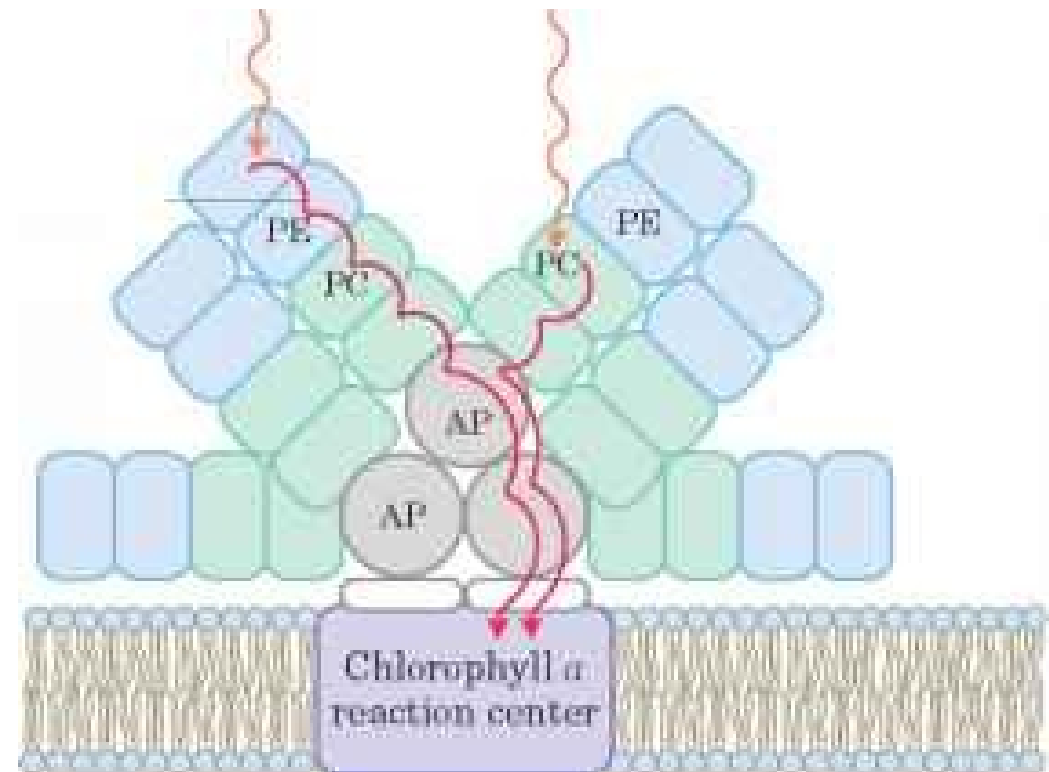
May 14, 2012, a few hours later ...



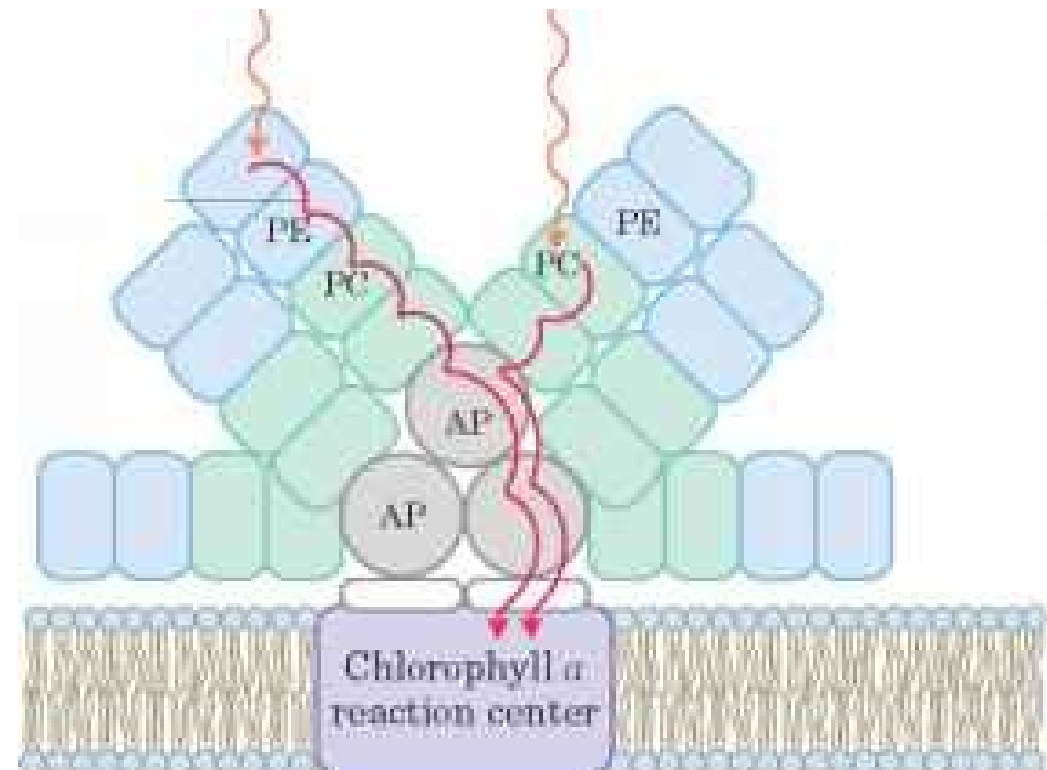
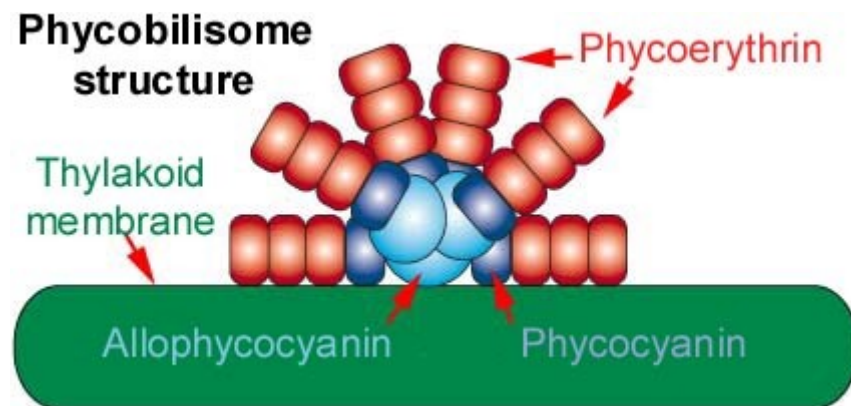
sunrise over pacific



May 14, 2012, and so Life begins ...

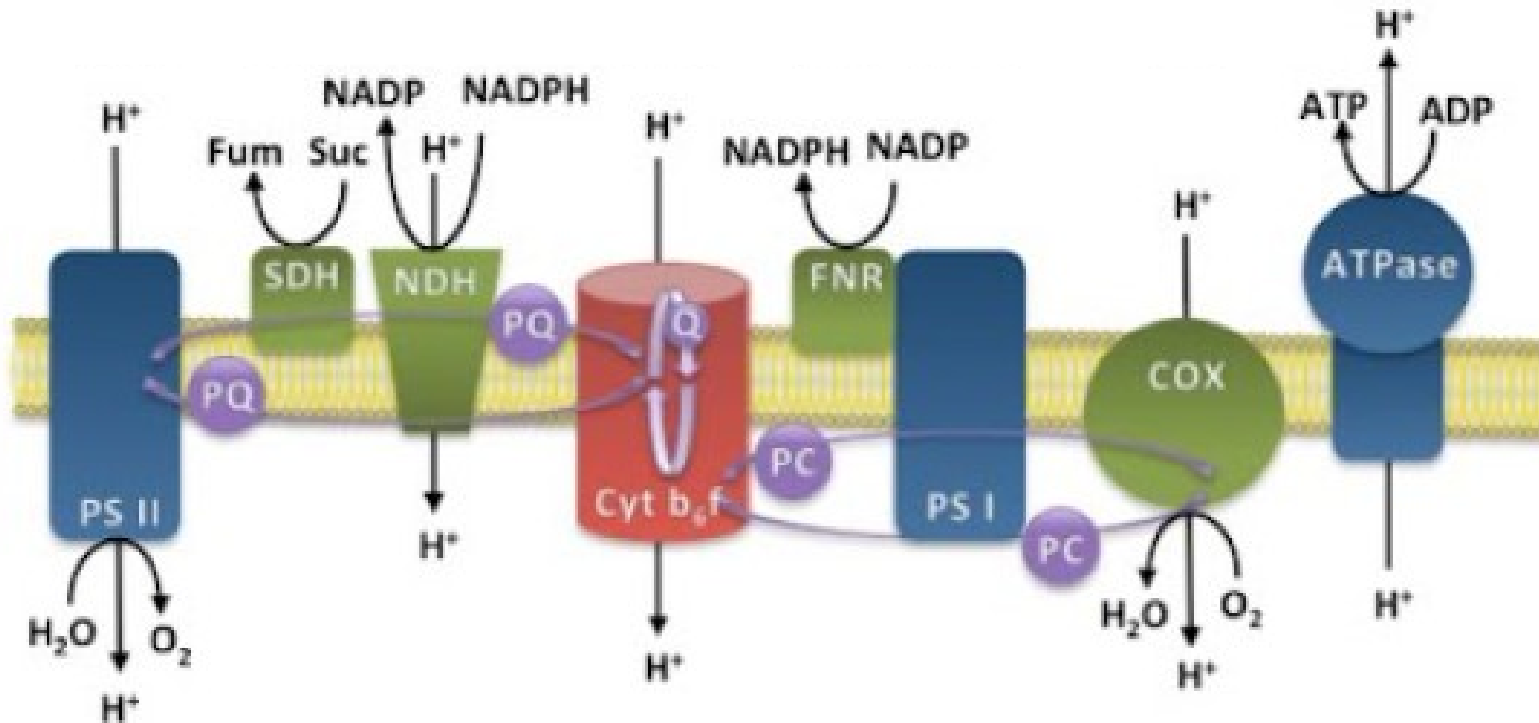


May 14, 2012, and so Life begins ...

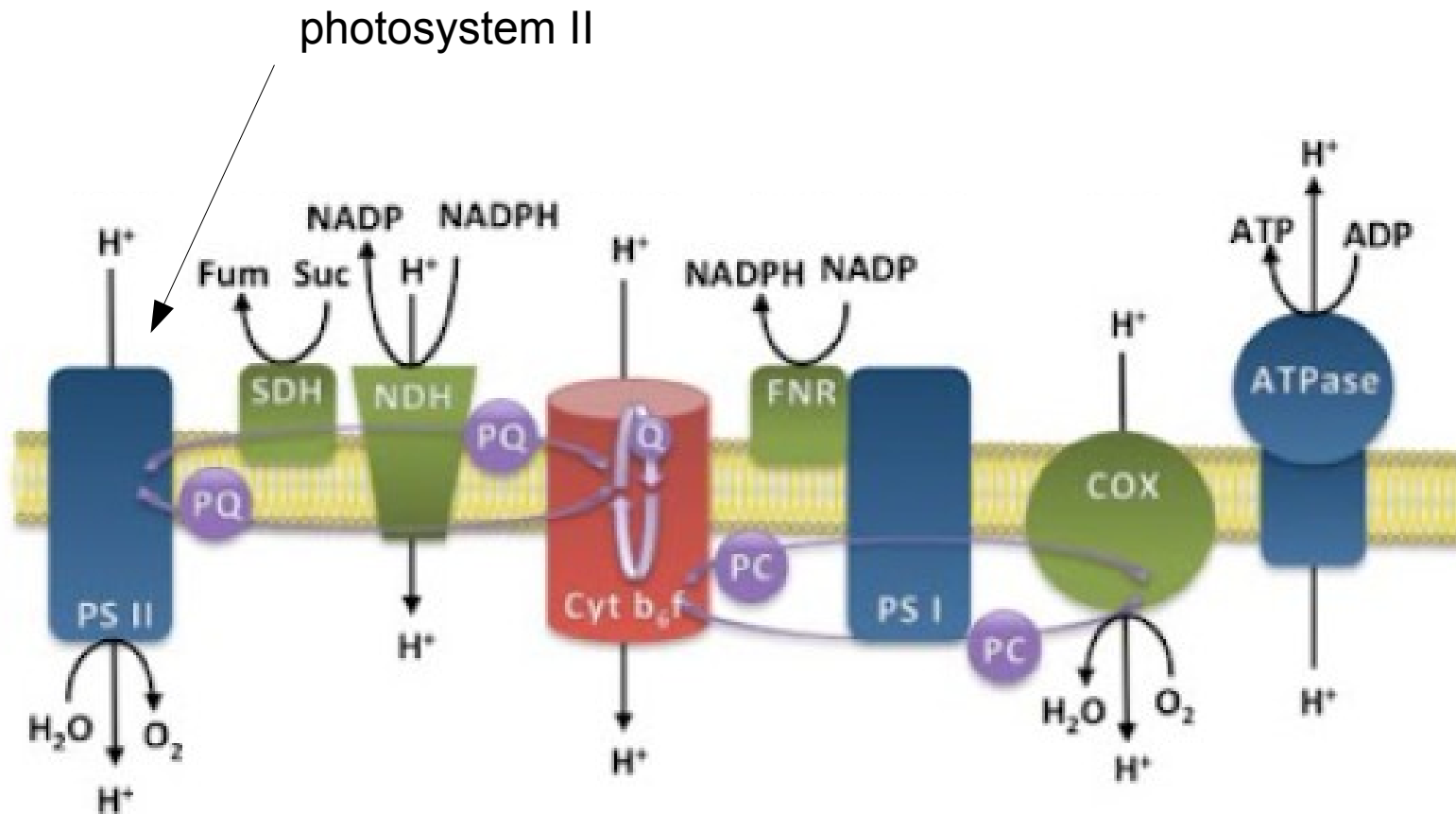




The light reactions

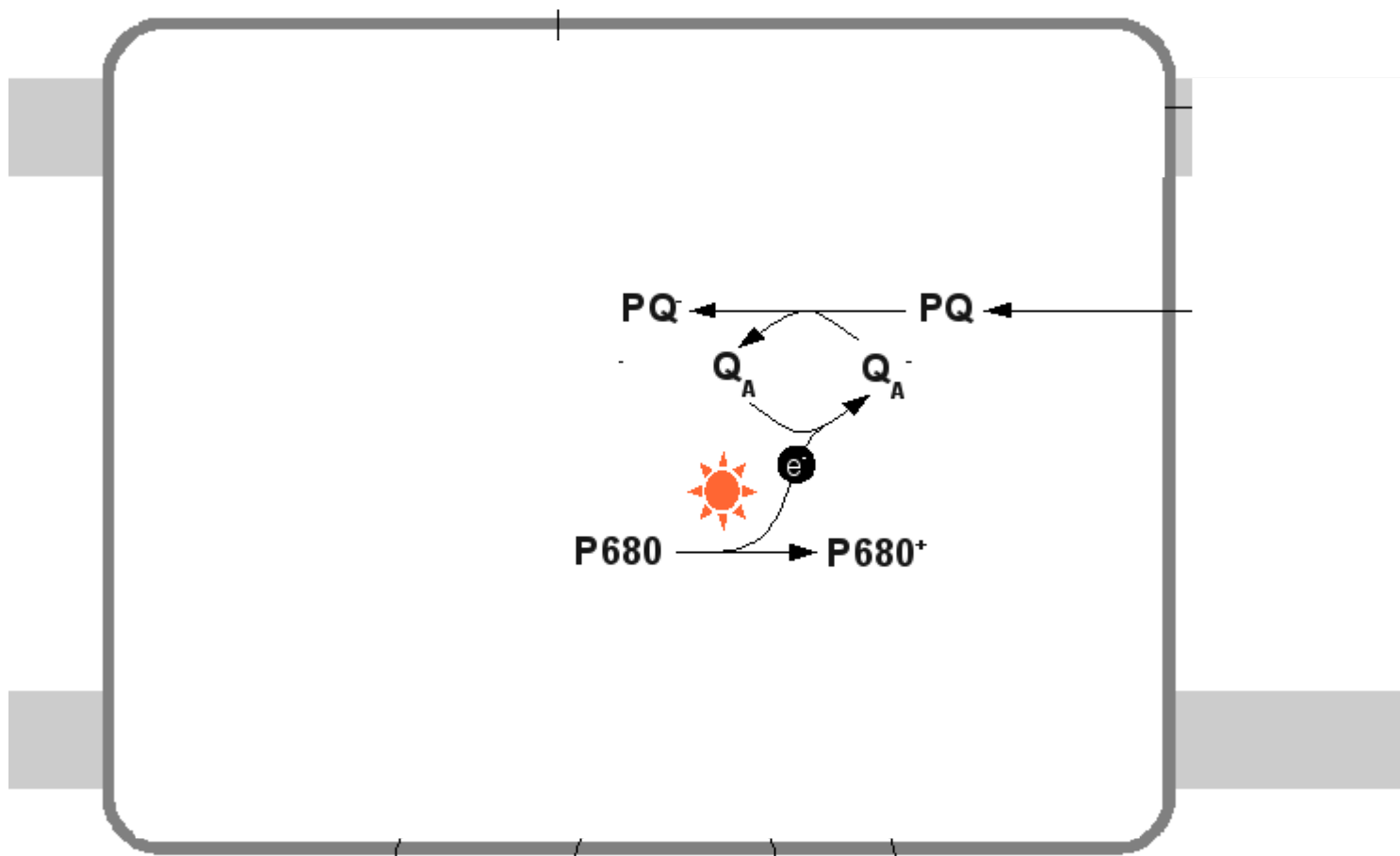


The light reactions

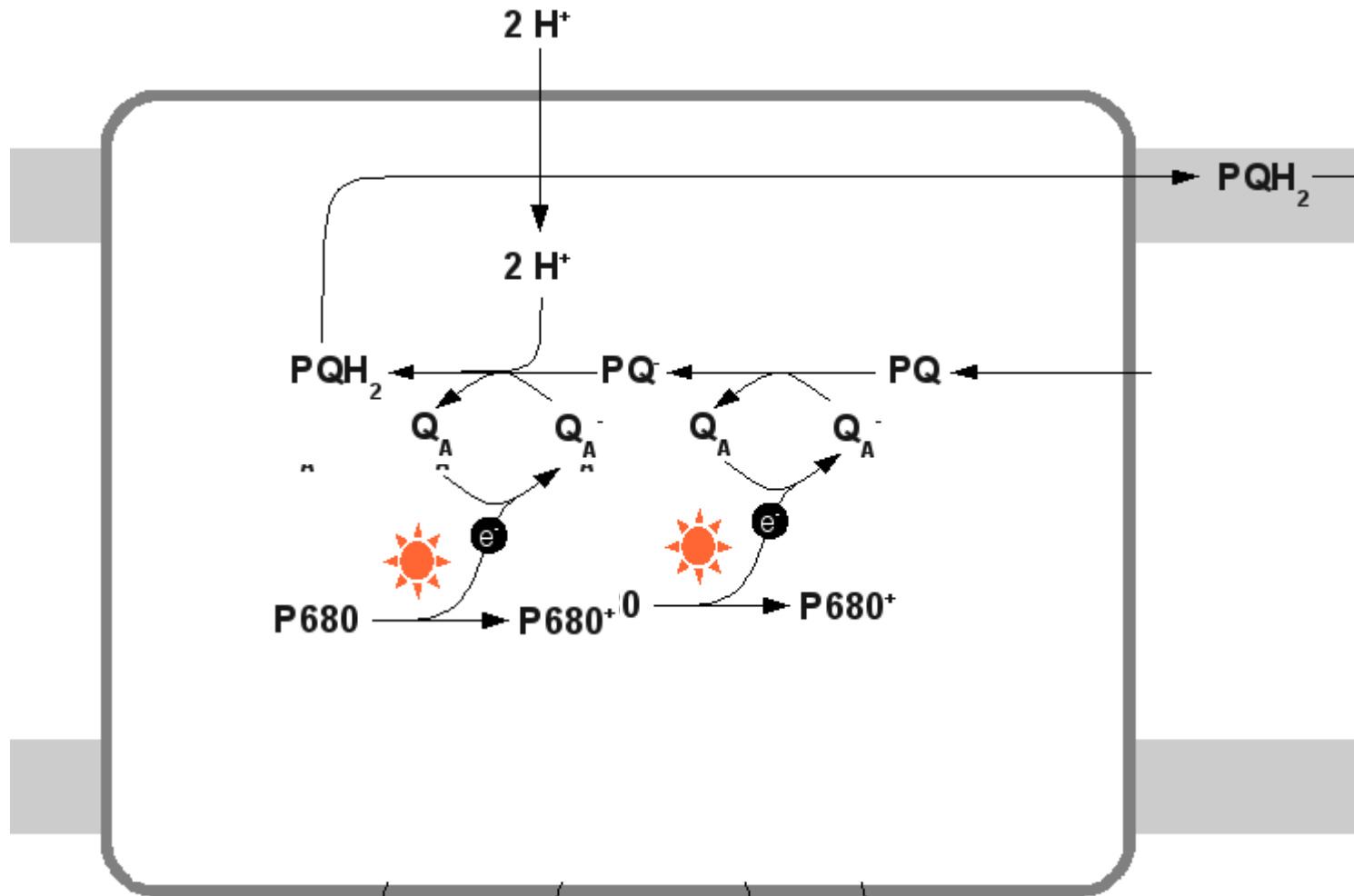




The light reactions: photosystem II

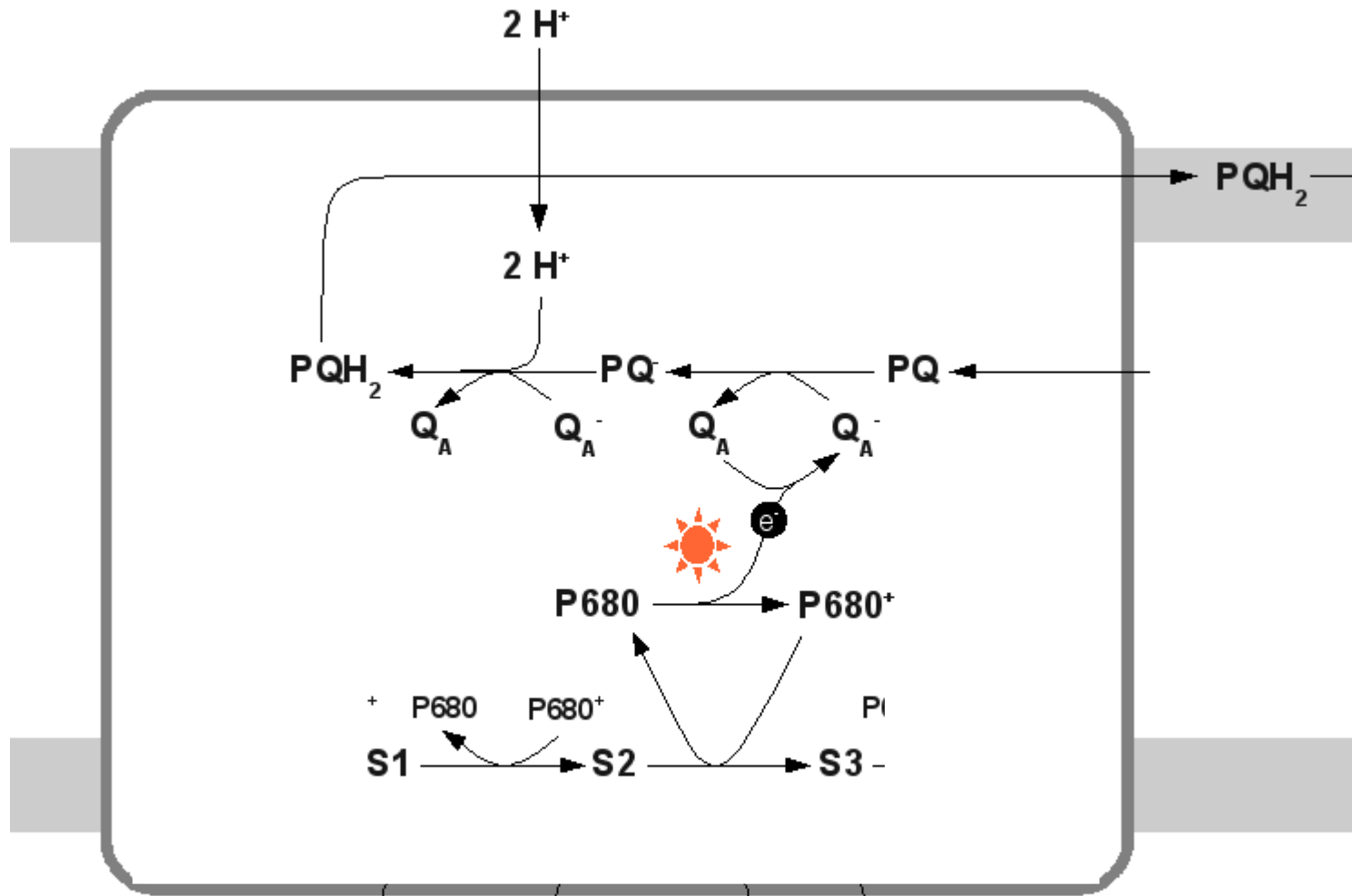


The light reactions: photosystem II

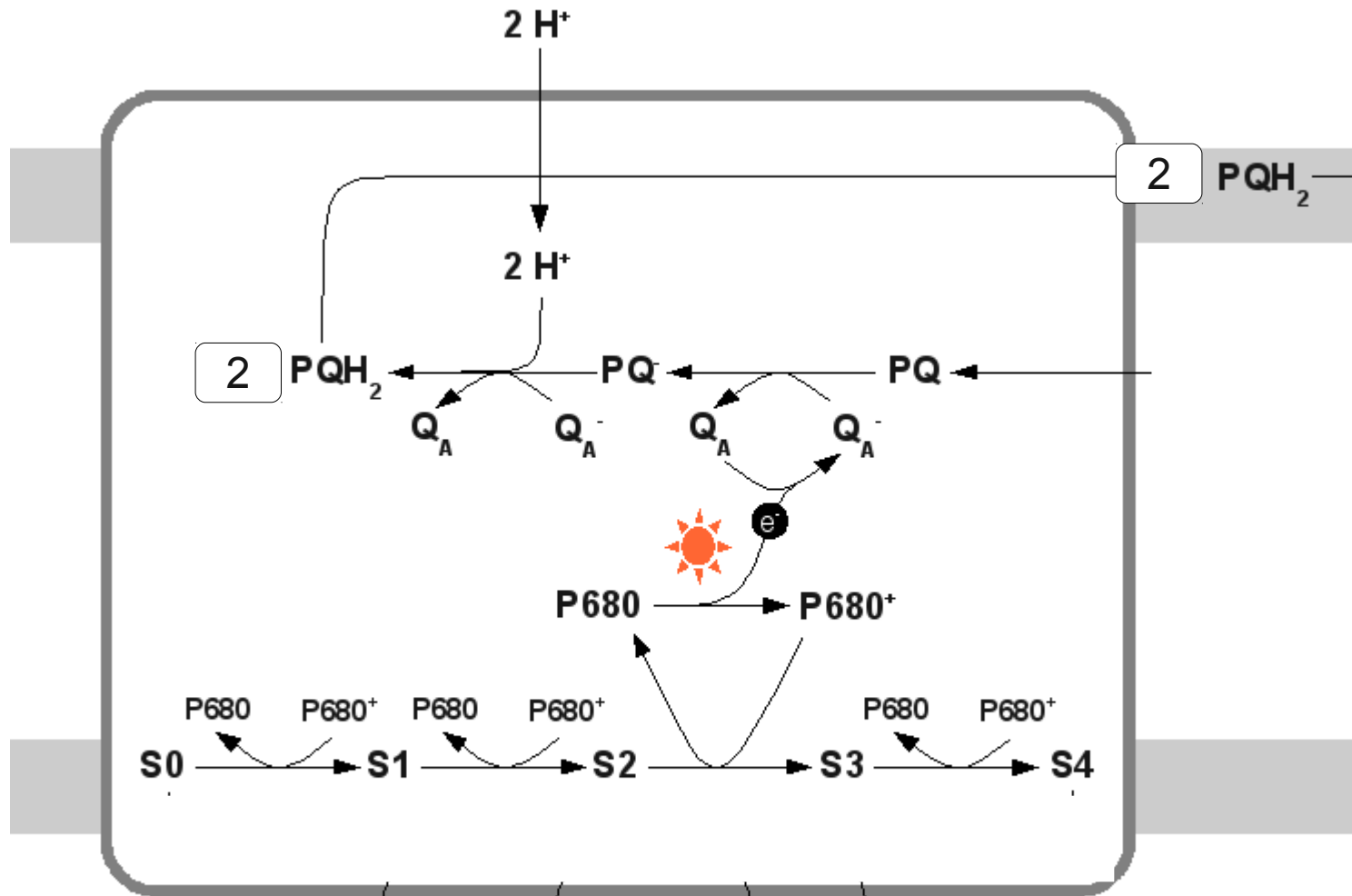




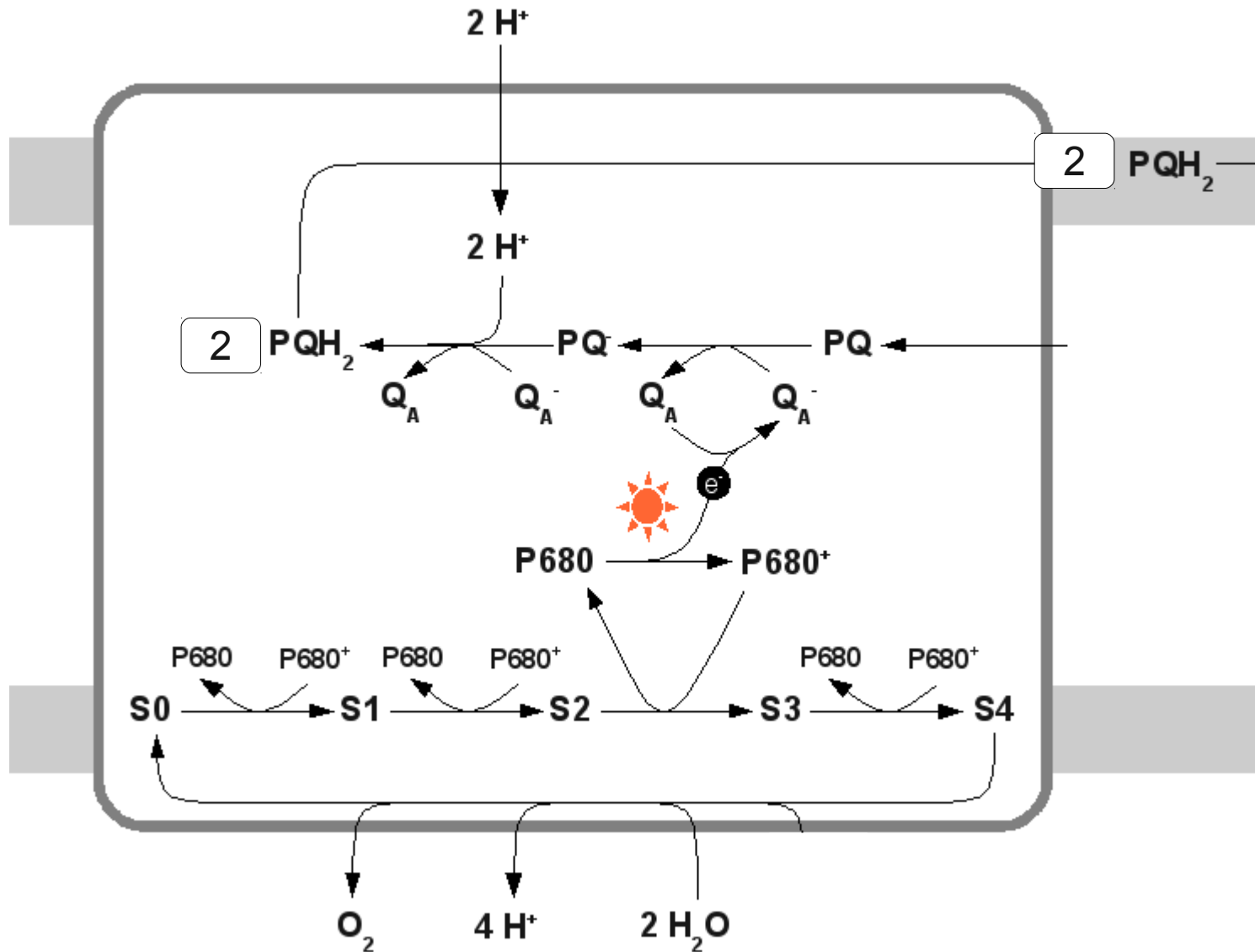
The light reactions: photosystem II



The light reactions: photosystem II

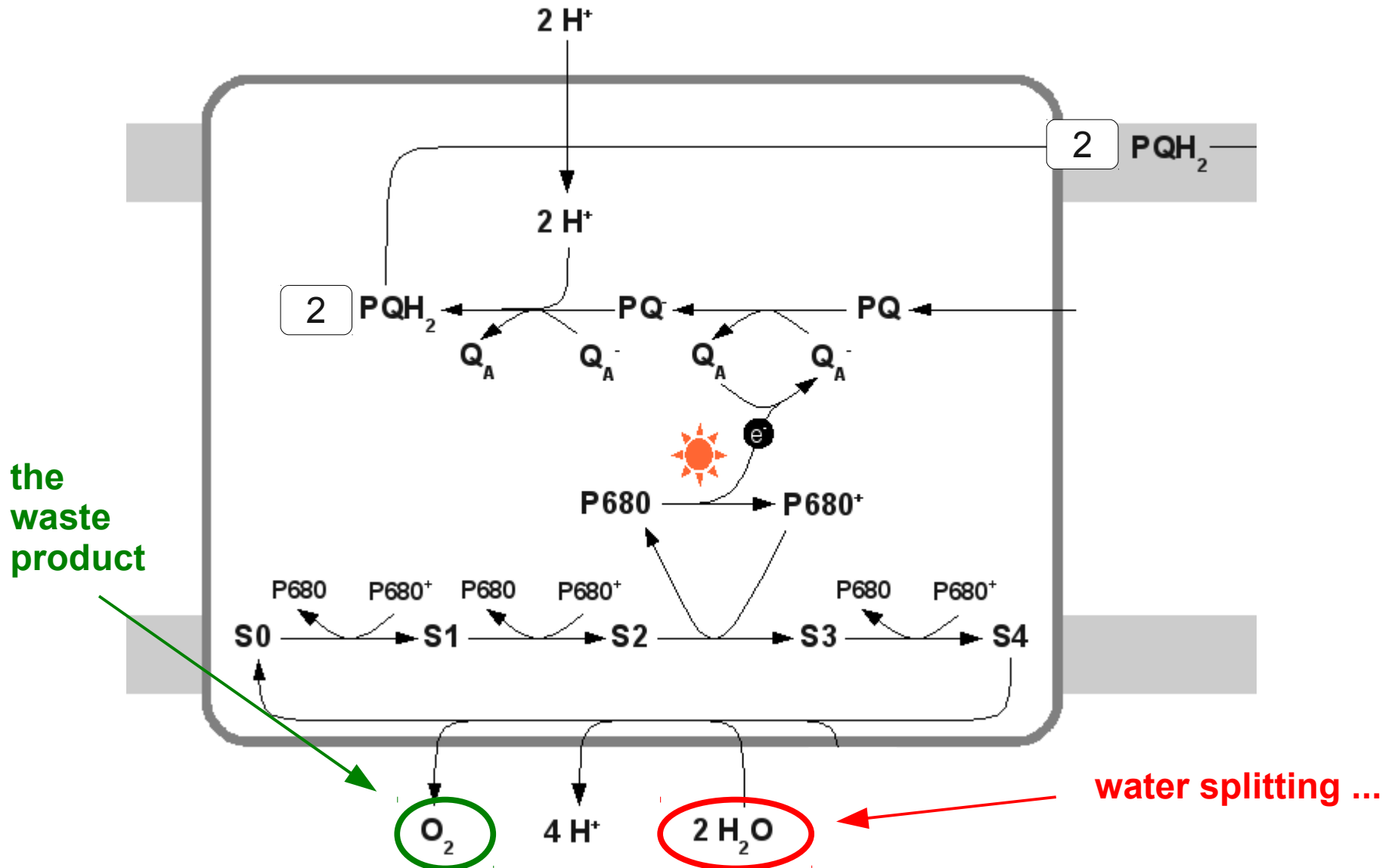


The light reactions: photosystem II

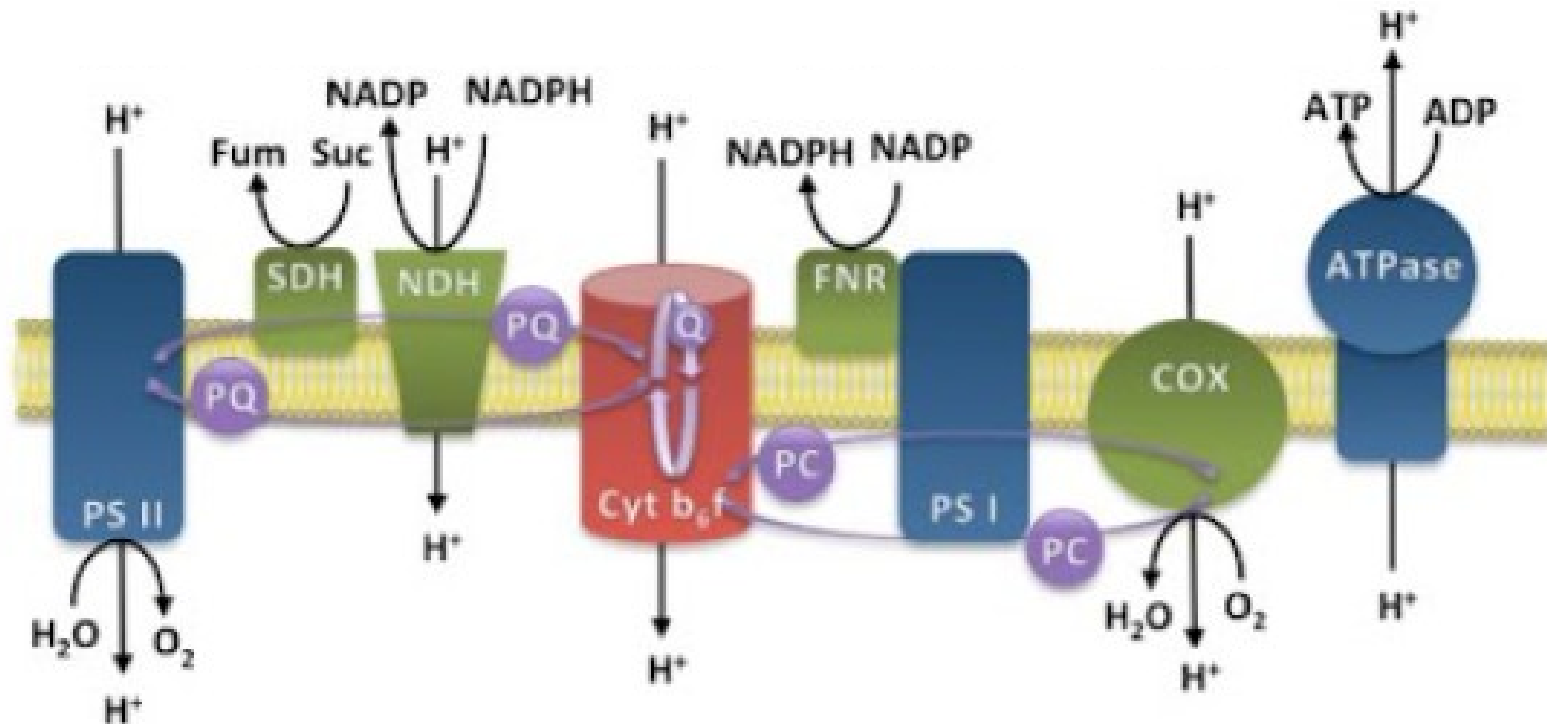




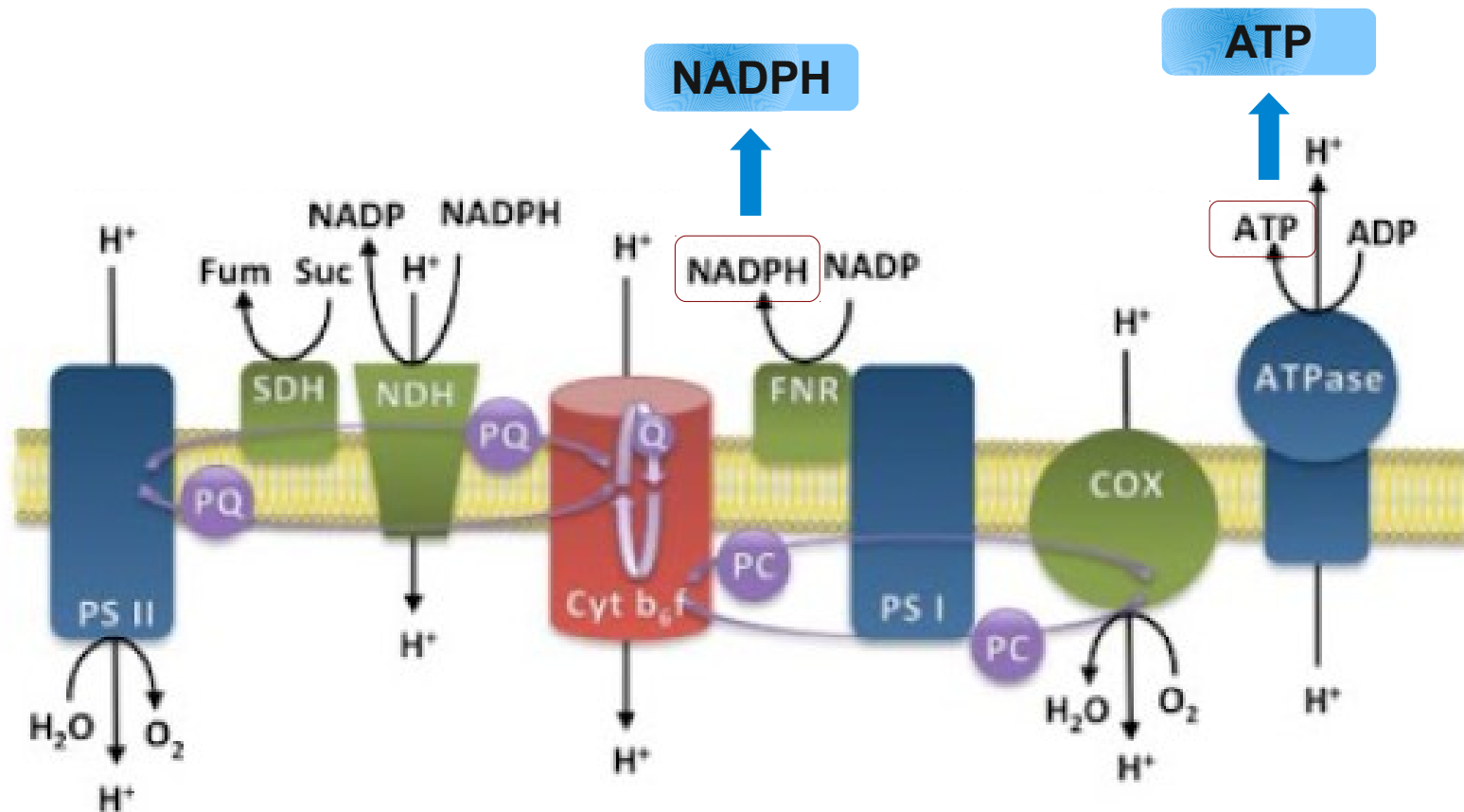
The light reactions: photosystem II



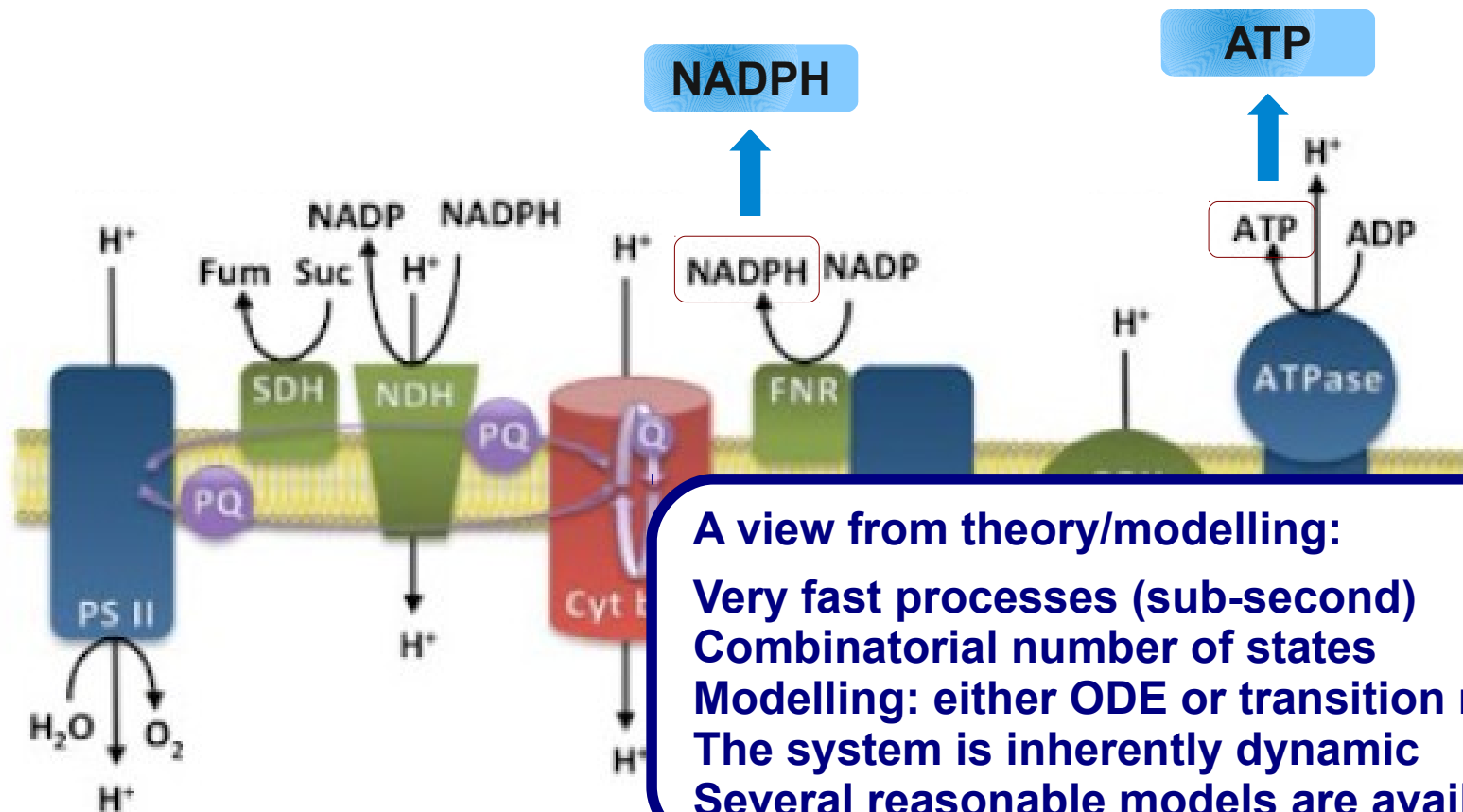
The light reactions: eating the sun



The light reactions: eating the sun



The light reactions: eating the sun



A view from theory/modelling:

Very fast processes (sub-second)

Combinatorial number of states

Modelling: either ODE or transition matrices

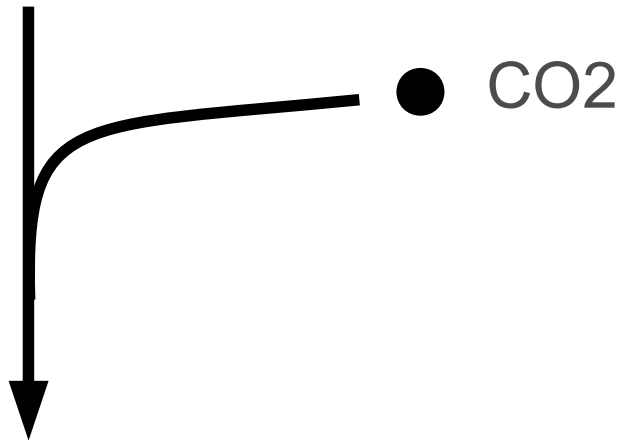
The system is inherently dynamic

Several reasonable models are available

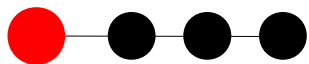


Fixation of atmospheric CO₂ by RuBisCO

 ribulose-1,5-bisphosphate (RuBP, 5 carbon)



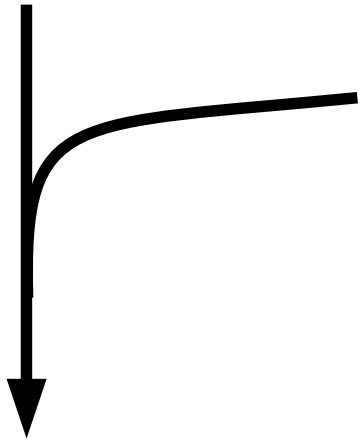
 3-phosphoglycerate (3 carbon)





Fixation of atmospheric CO₂ by RuBisCO

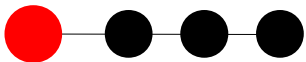
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 C

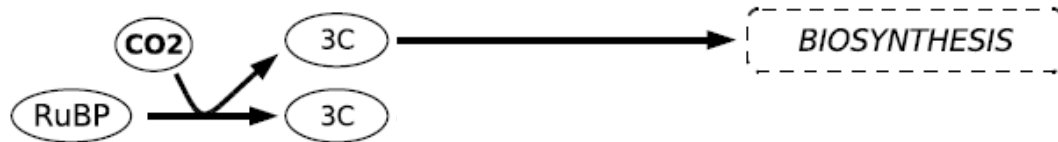
A view from theory/modelling:
RubisCo is slow and sloppy
Only few interconversions per second
A limiting factor in phototrophic growth.
Low specificity to its substrate
Modelling: usually ODE/enzyme kinetics

 3-phosphoglycerate (3 carbon)



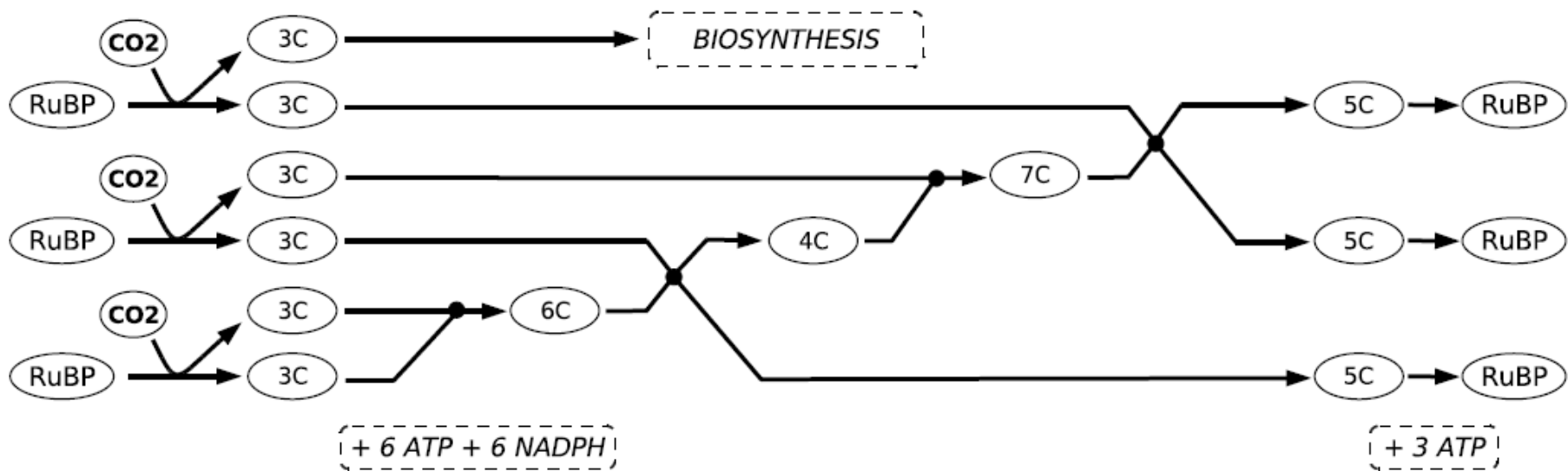


The Calvin-Benson-Bassham (CBB) cycle: Regeneration





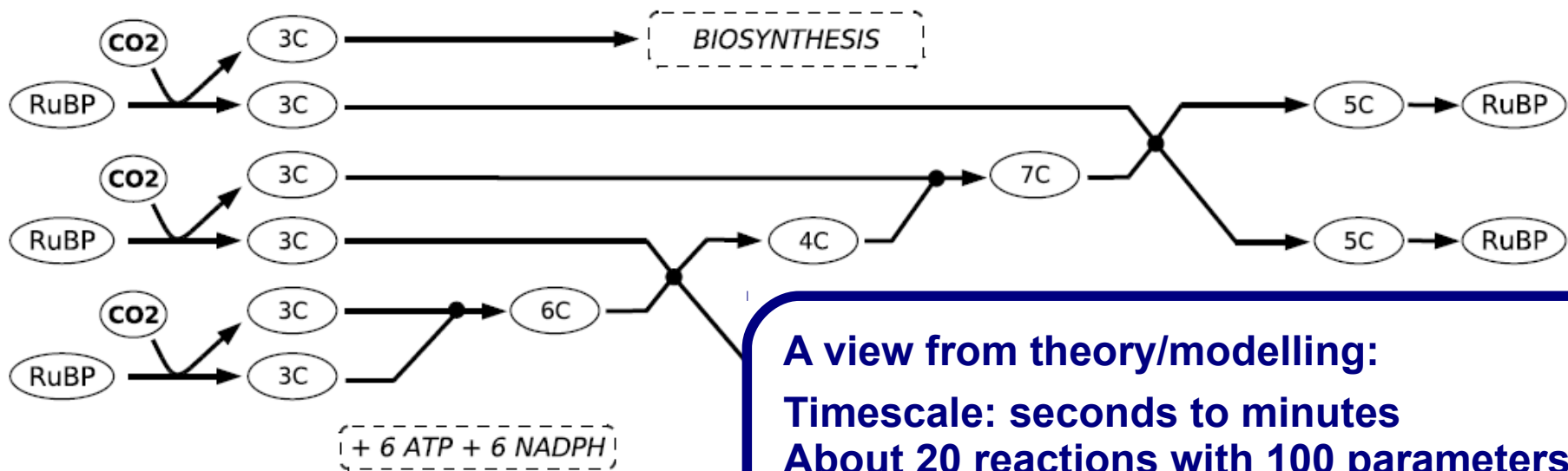
The Calvin-Benson-Bassham (CBB) cycle: Regeneration



From: R. Steuer and B. H. Junker. (2009) Computational Models of Metabolism: Stability and Regulation in Metabolic Networks. Advances in Chemical Physics, Volume 142



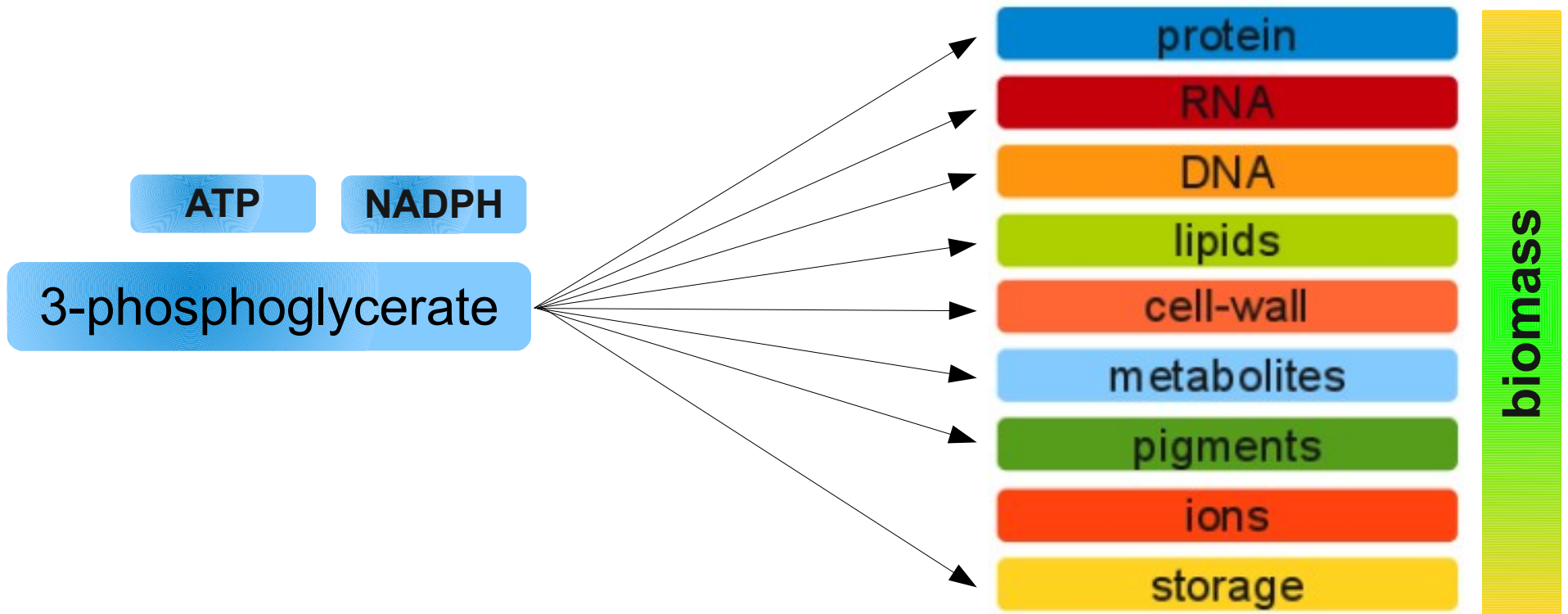
The Calvin-Benson-Bassham (CBB) cycle: Regeneration



A view from theory/modelling:
Timescale: seconds to minutes
About 20 reactions with 100 parameters.
Typically implemented as ODE model
Q: are there alternative cycles?

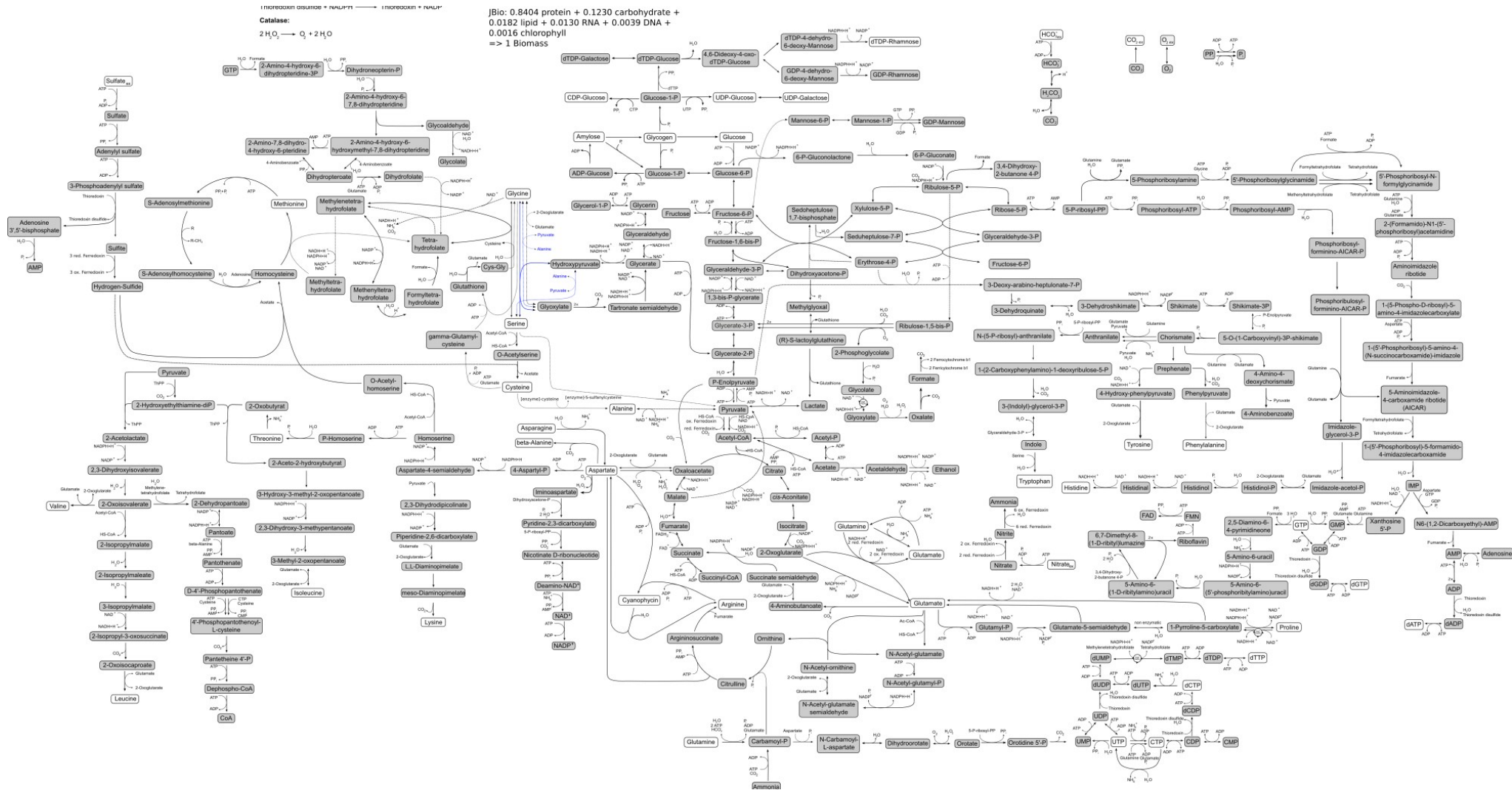


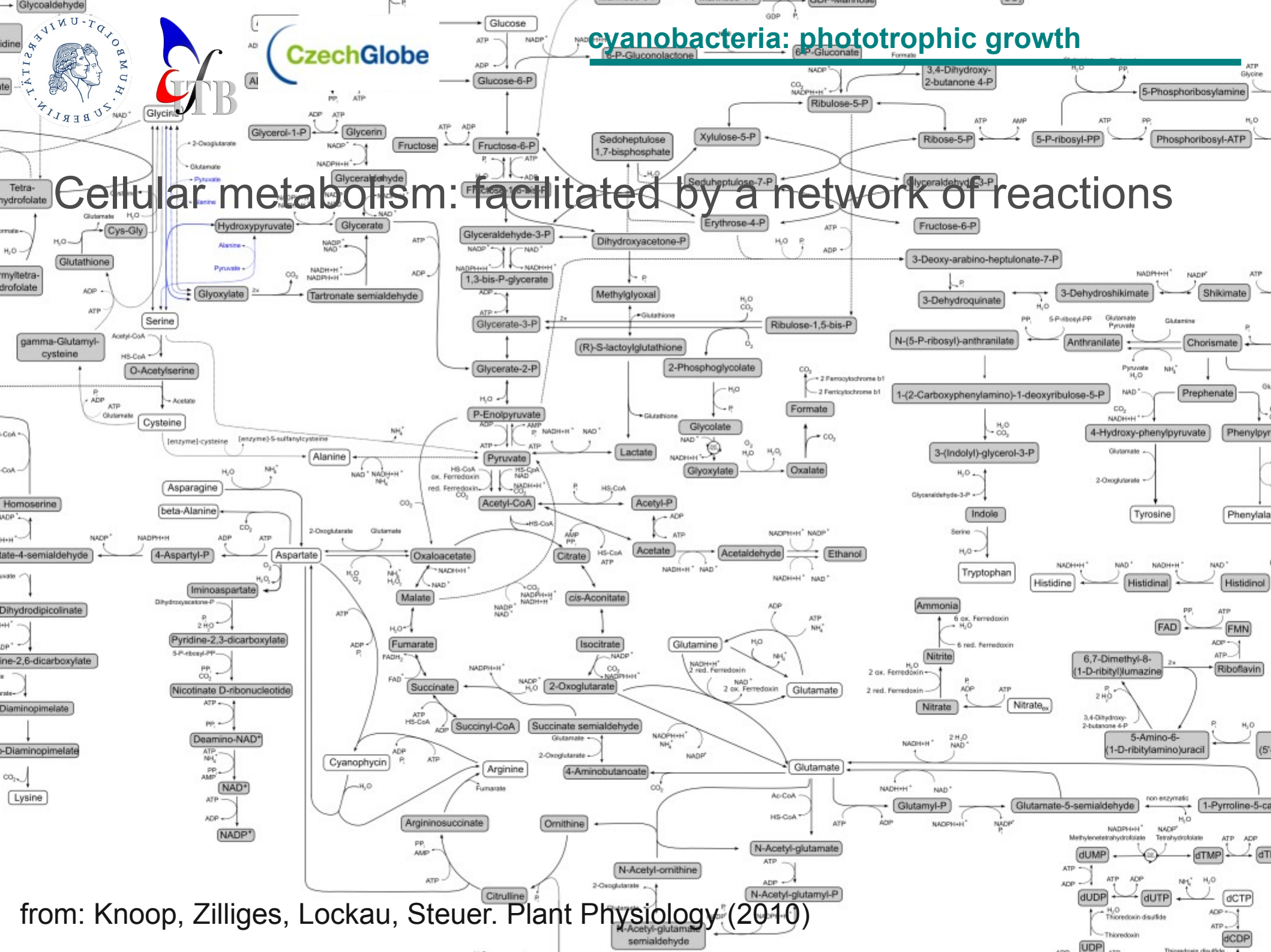
The Calvin-Benson-Bassham (CBB) cycle: Regeneration





Cellular metabolism: facilitated by a network of reactions





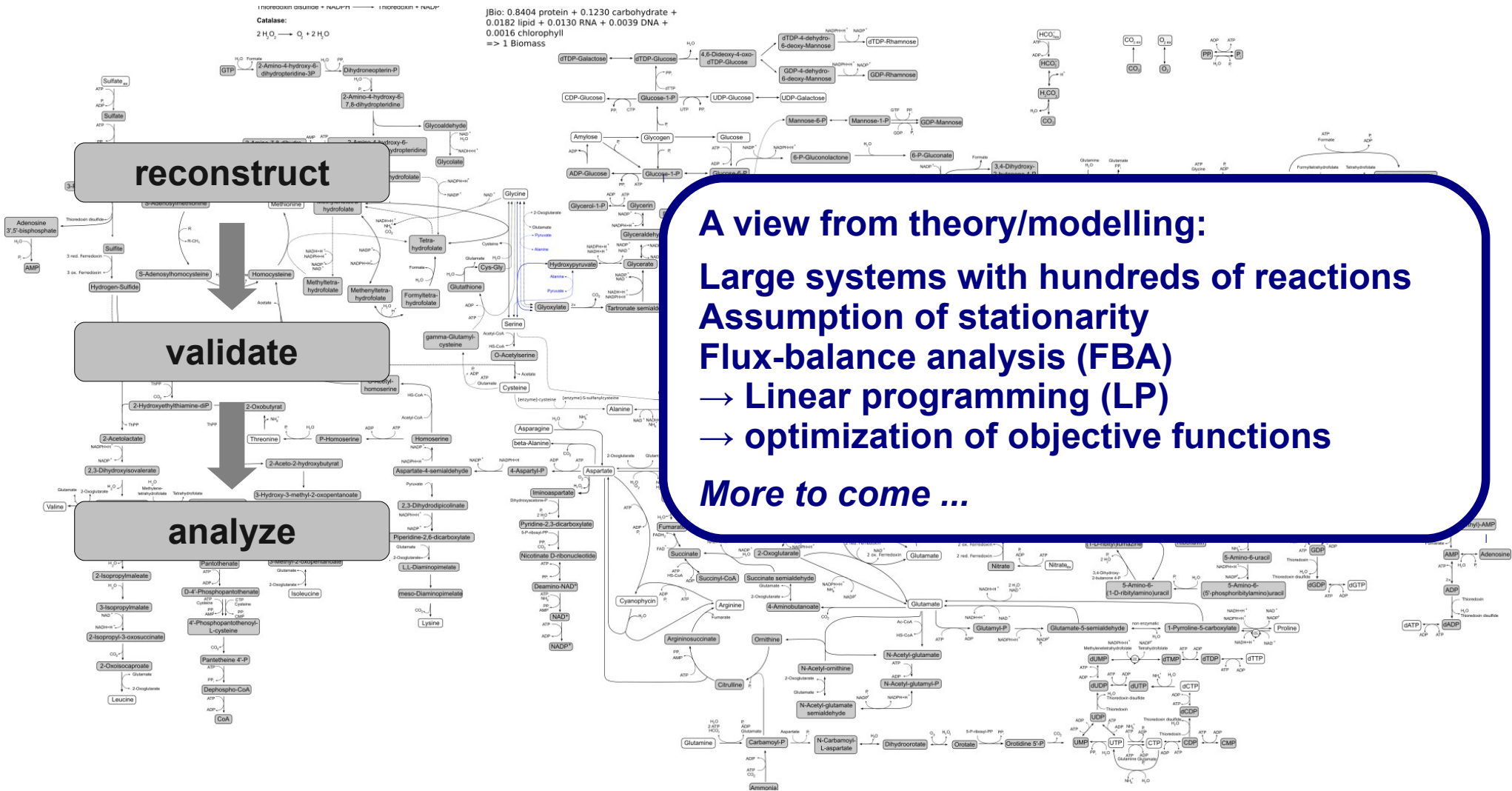
cyanobacteria: phototrophic growth

Cellular metabolism: facilitated by a network of reactions

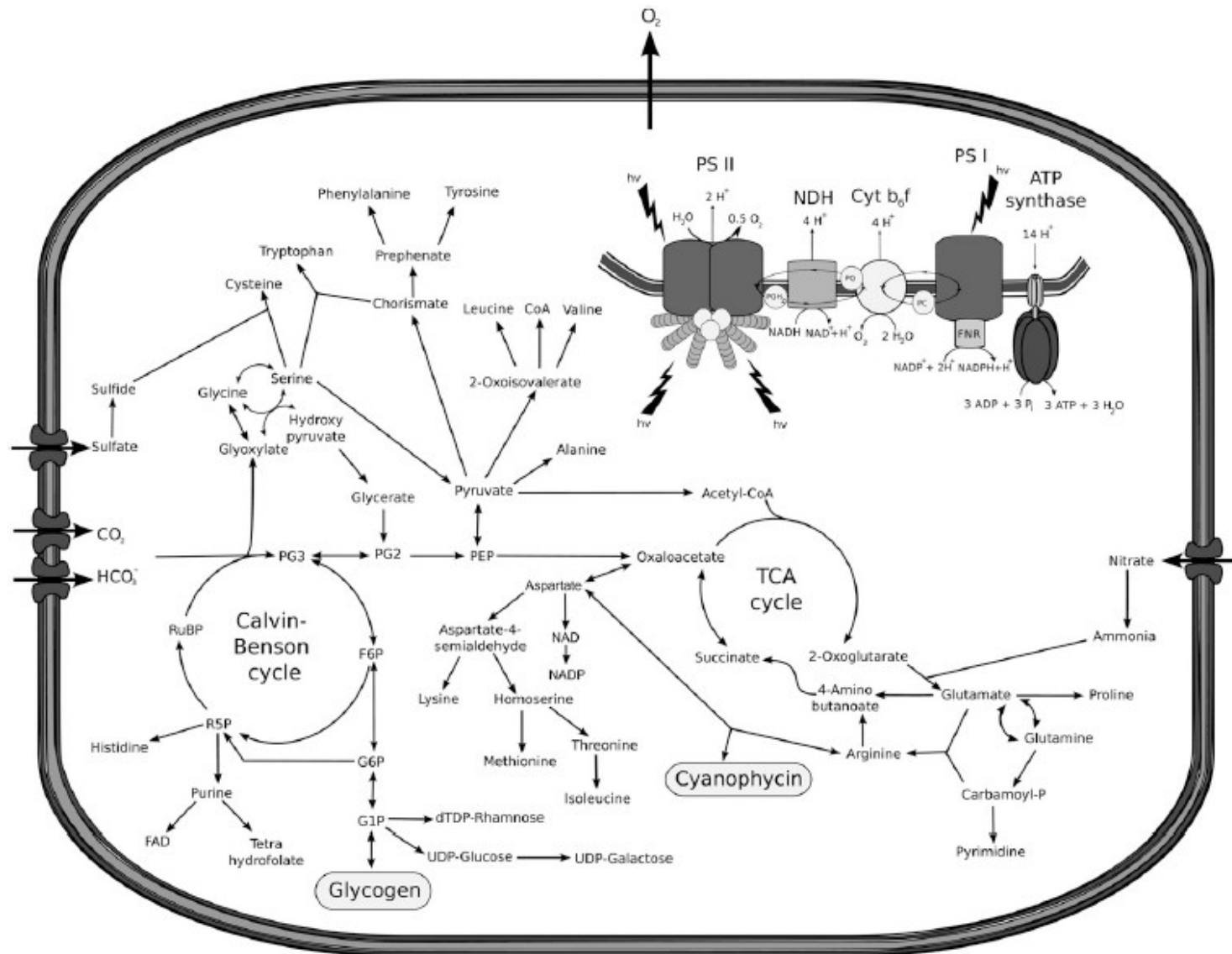
from: Knoop, Zilliges, Lockau, Steuer. Plant Physiology (2010)



Cellular metabolism: facilitated by a network of reactions

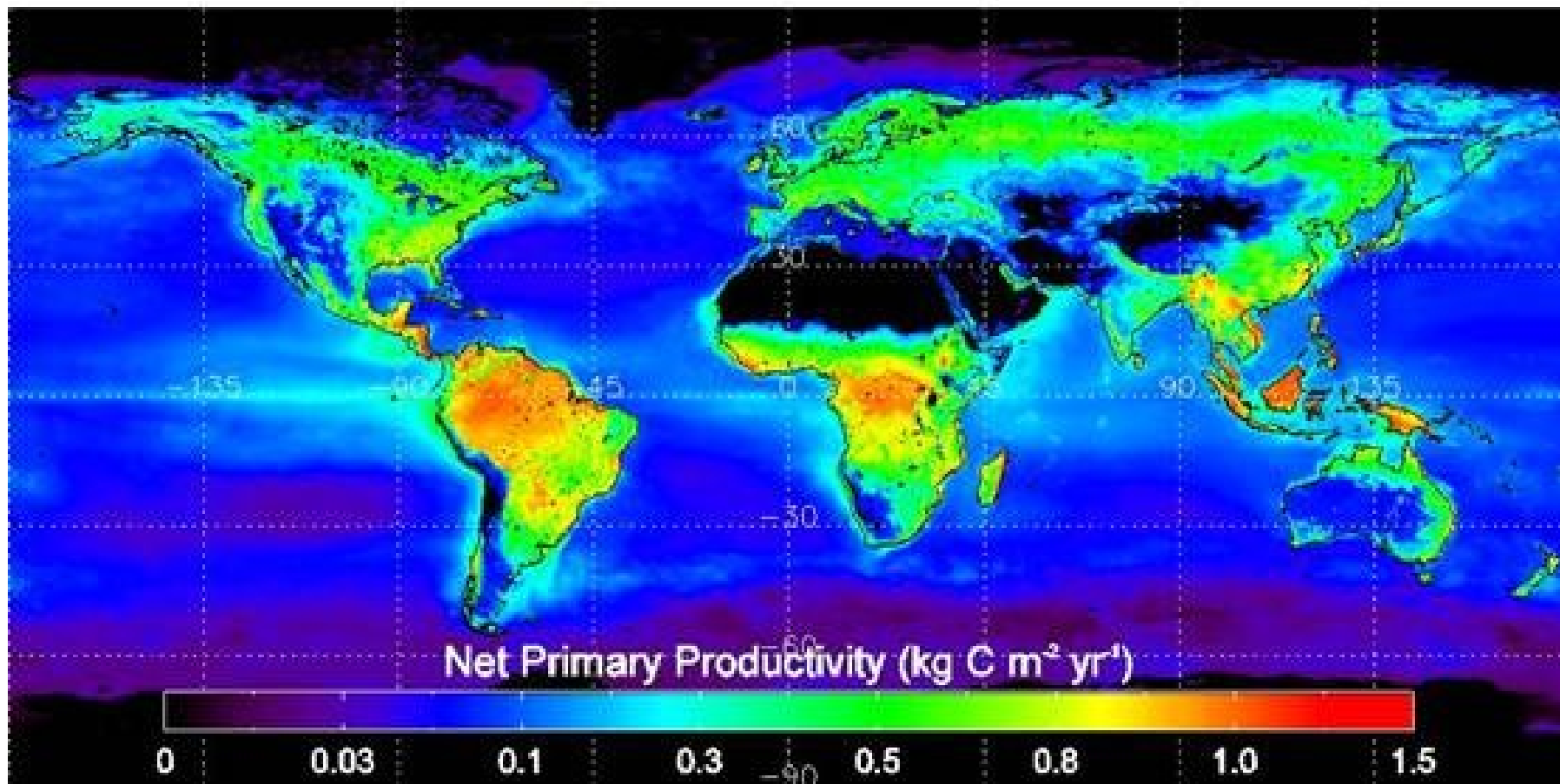


Cellular metabolism: facilitated by a network of reactions





Phototrophic growth and the environment:





Phototrophic growth and the environment:

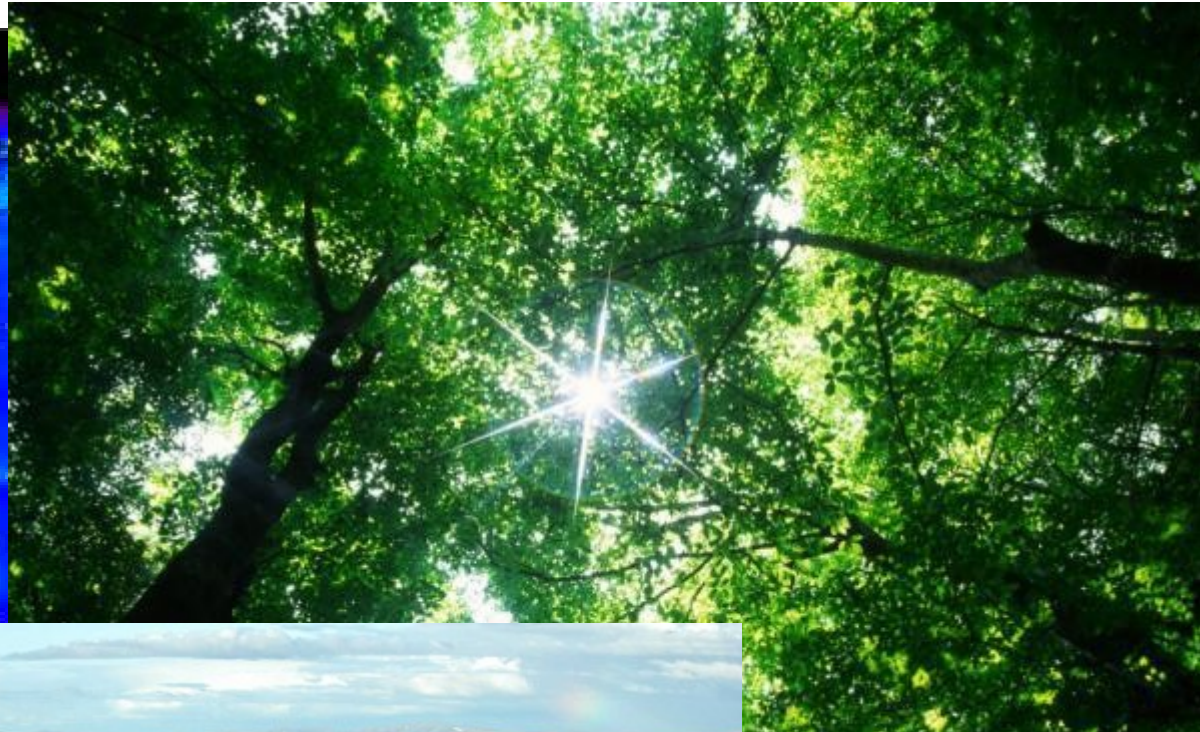
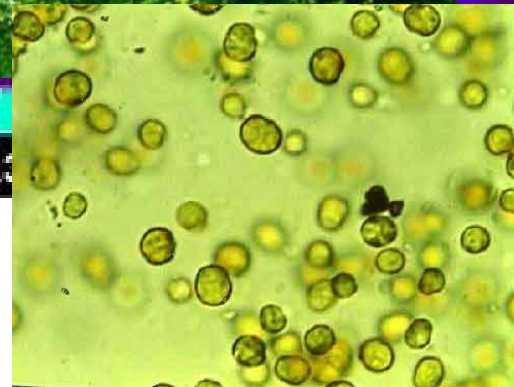


photo by Daniel R. Ruthrauff, USGS





May 14, 2012, 11:59pm, by the end of the day ...



1.1×10^{19} joule solar energy is absorbed by Earth's atmosphere, oceans and land masses per day ...

600 000 000 tons of carbon fixed by photosynthesis

120 Gt carbon per year (land)

90 Gt carbon per year (ocean)



Cyanobacteria: a hierarchy of processes

We aim to understand the life and growth of cyanobacteria





Cyanobacteria: a hierarchy of processes

We aim to understand the life and growth of cyanobacteria



The CyanoTeam and CyanoNetwork

An association between several groups from EU, Israel, and USA to model and understand a cyanobacterial cell *in silico*.

International team led by John Whitmarsh
Coordinator local experimental team: Ladislav Nedbal
Coordinator local modelling team: Ralf Steuer





Cyanobacteria: a hierarchy of processes

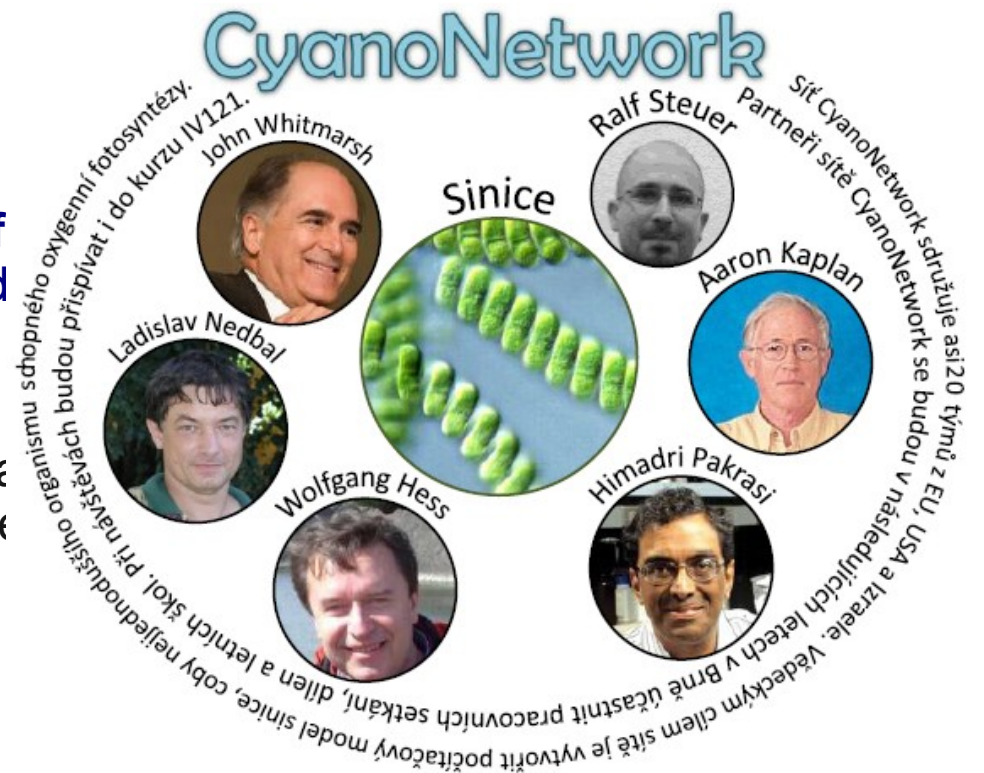
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The CyanoTeam and CyanoNetwork

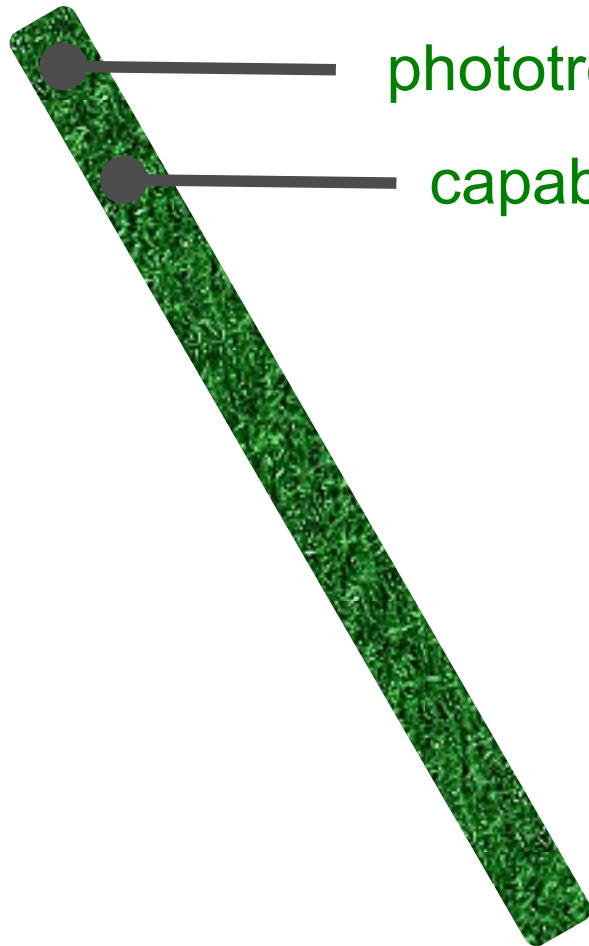
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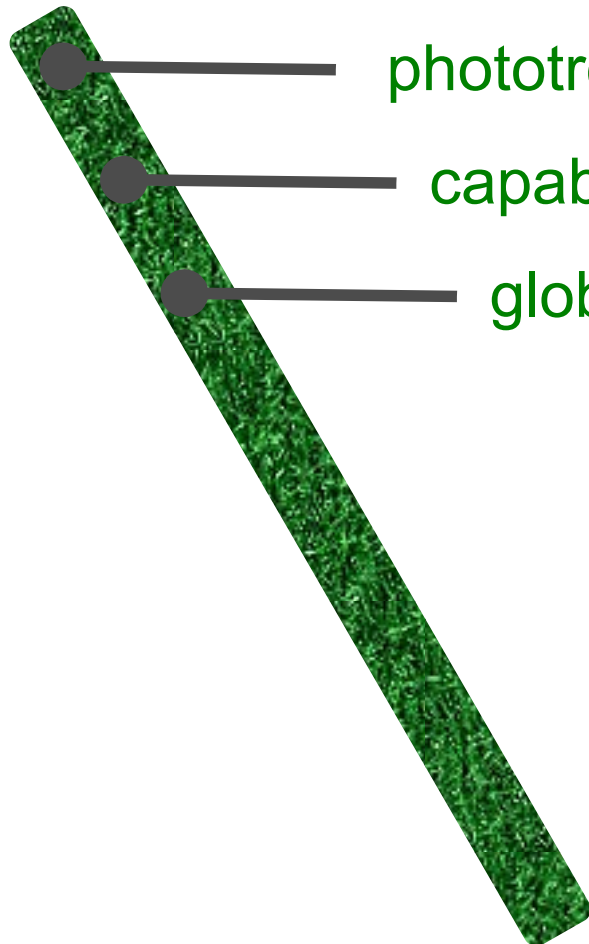
Cyanobacteria: understanding phototrophic growth



phototrophic micro-organisms (prokaryotes)

capable of oxygen-evolving photosynthesis

Cyanobacteria: understanding phototrophic growth



phototrophic micro-organisms (prokaryotes)

capable of oxygen-evolving photosynthesis

globally extremely abundant

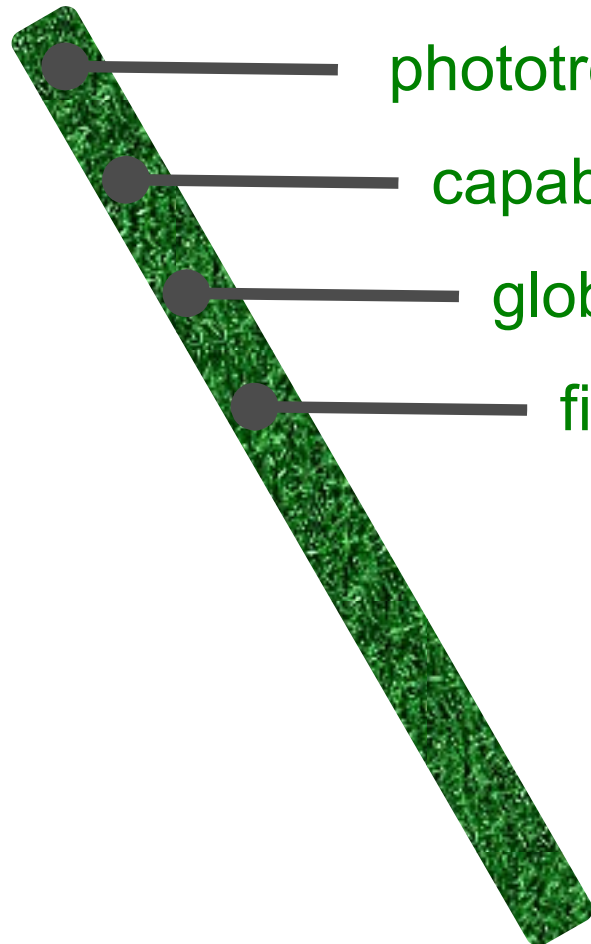
The cyanobacterium *Prochlorococcus* is the numerically dominant phototroph in some oceans (up to half of the photosynthetic biomass).

Cyanobacterial abundance in stratified oligotrophic waters can be high ($> 10^5$ cells ml^{-1})

from: Sullivan et al. *Nature* (2003)



Cyanobacteria: understanding phototrophic growth



phototrophic micro-organisms (prokaryotes)

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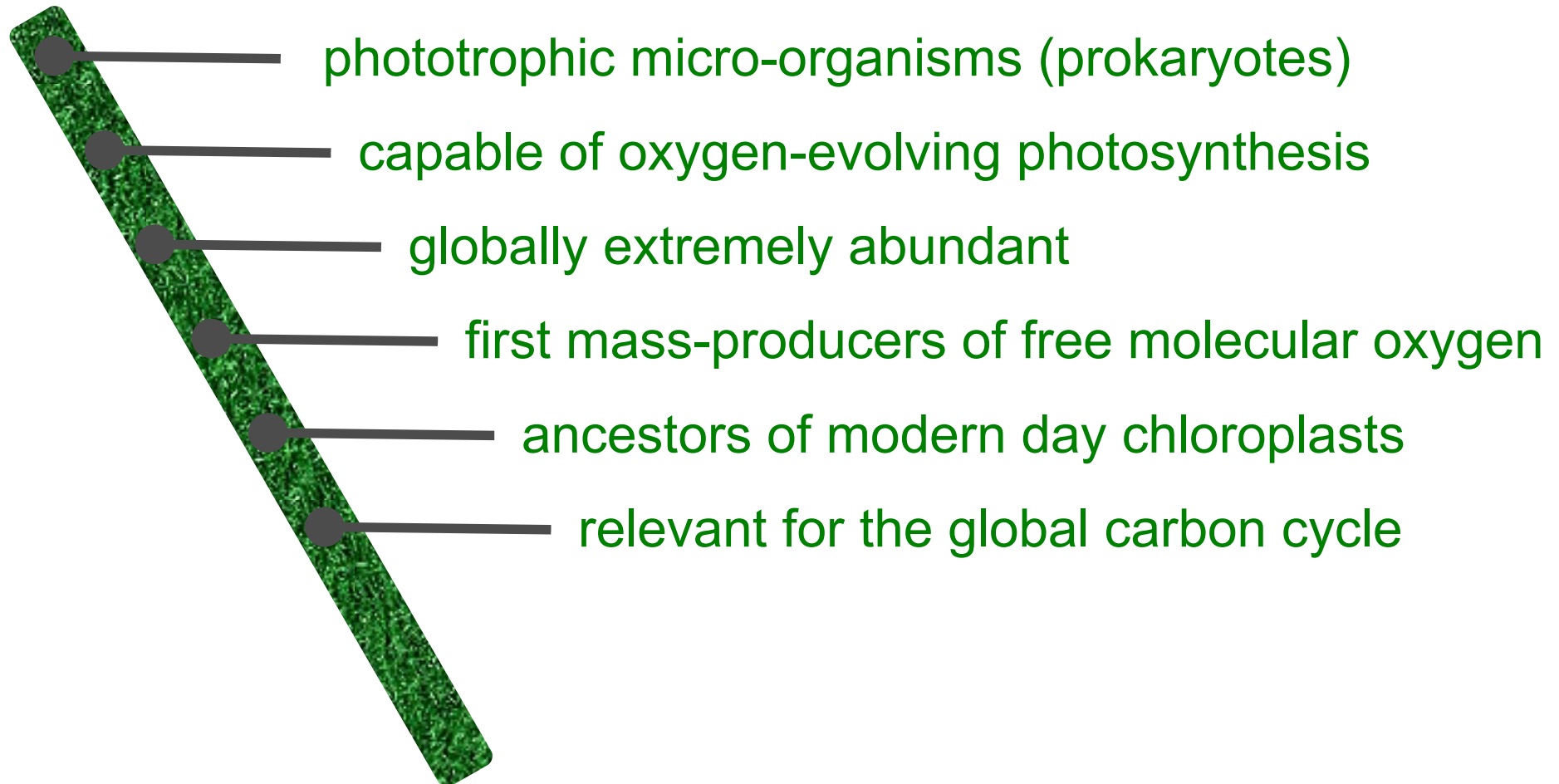
first mass-producers of free molecular oxygen



responsible for the Great Oxygenation Event (GOE) around 2.4 billion years ago.

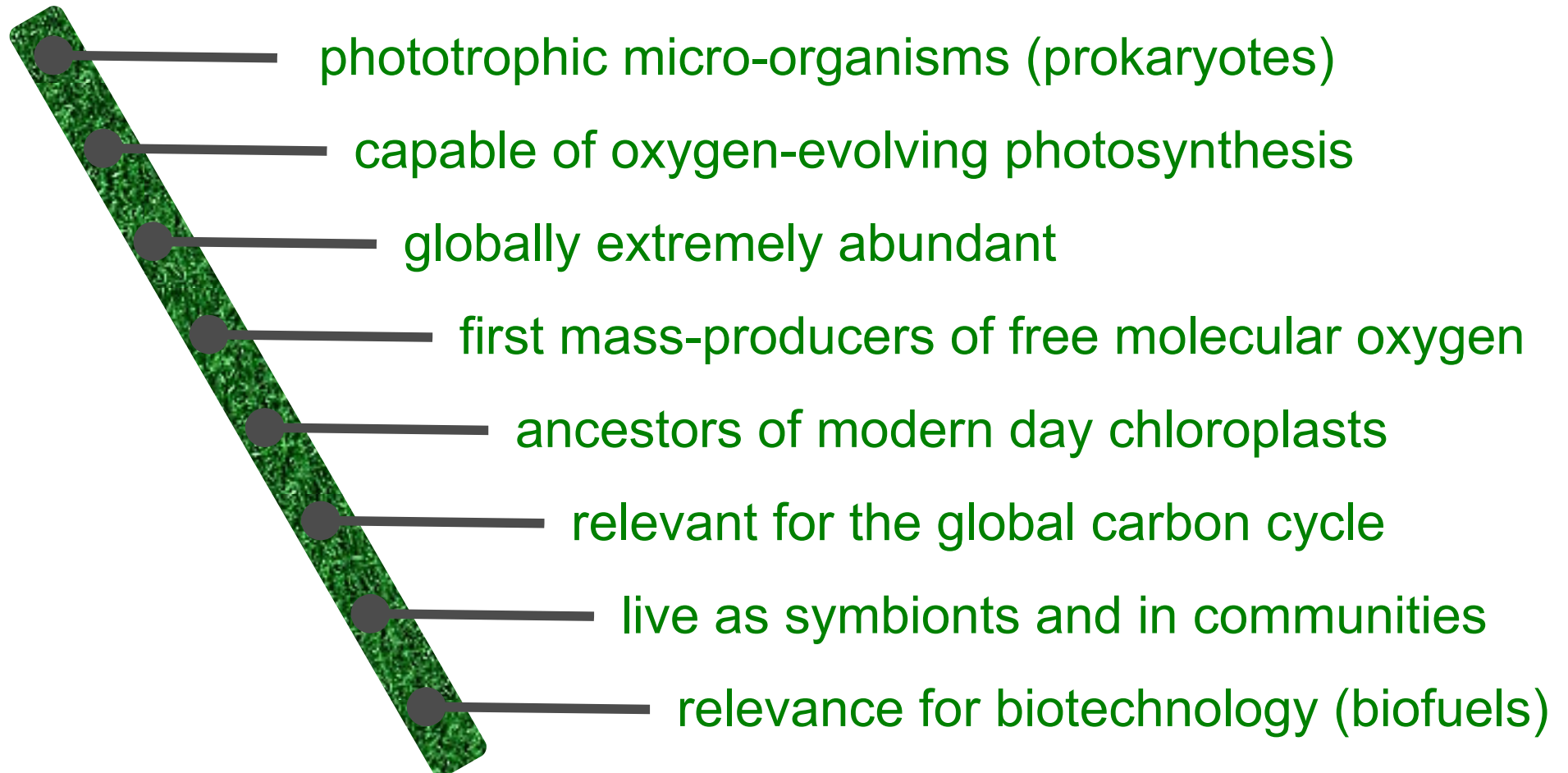


Cyanobacteria: understanding phototrophic growth



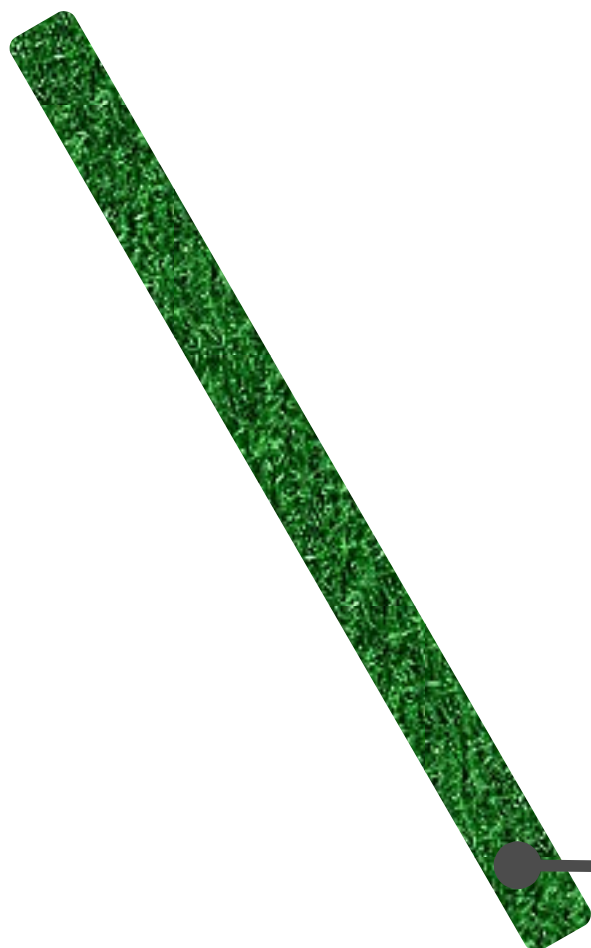


Cyanobacteria: understanding phototrophic growth





Cyanobacteria: understanding phototrophic growth



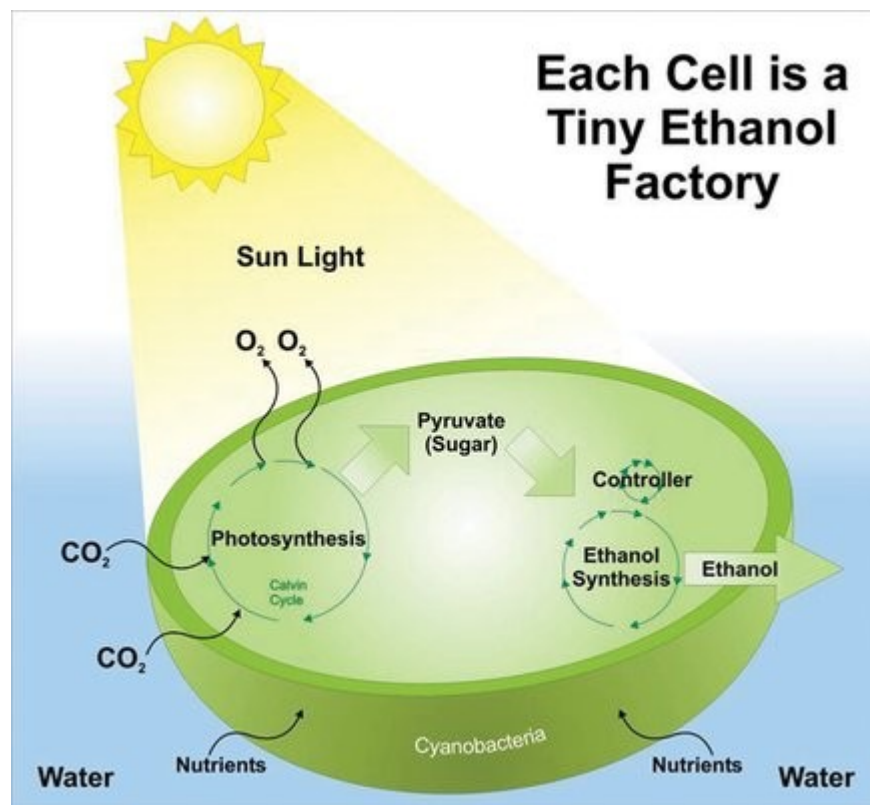
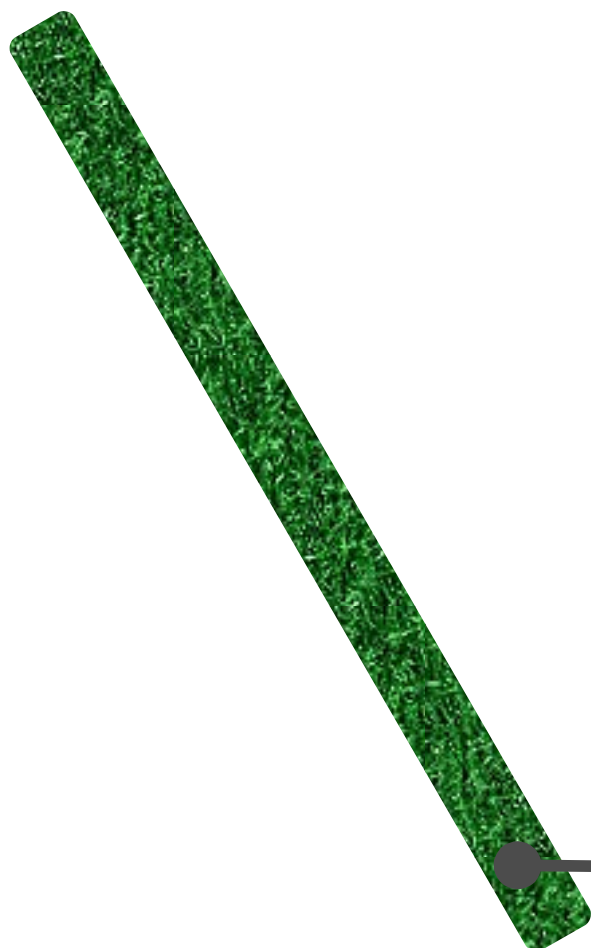
An open pond Spirulina farm:



relevance for biotechnology (biofuels)



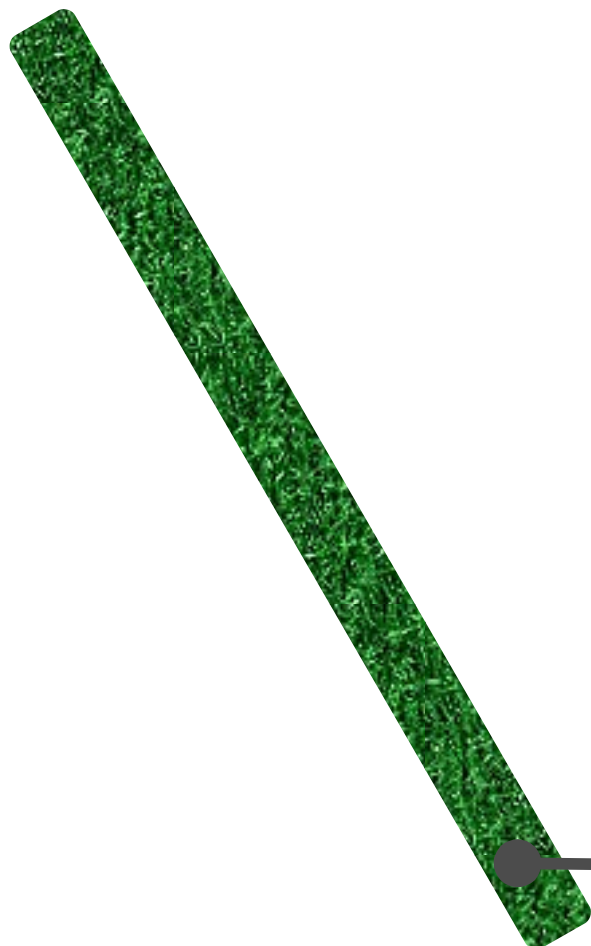
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Cyanobacteria: understanding phototrophic growth



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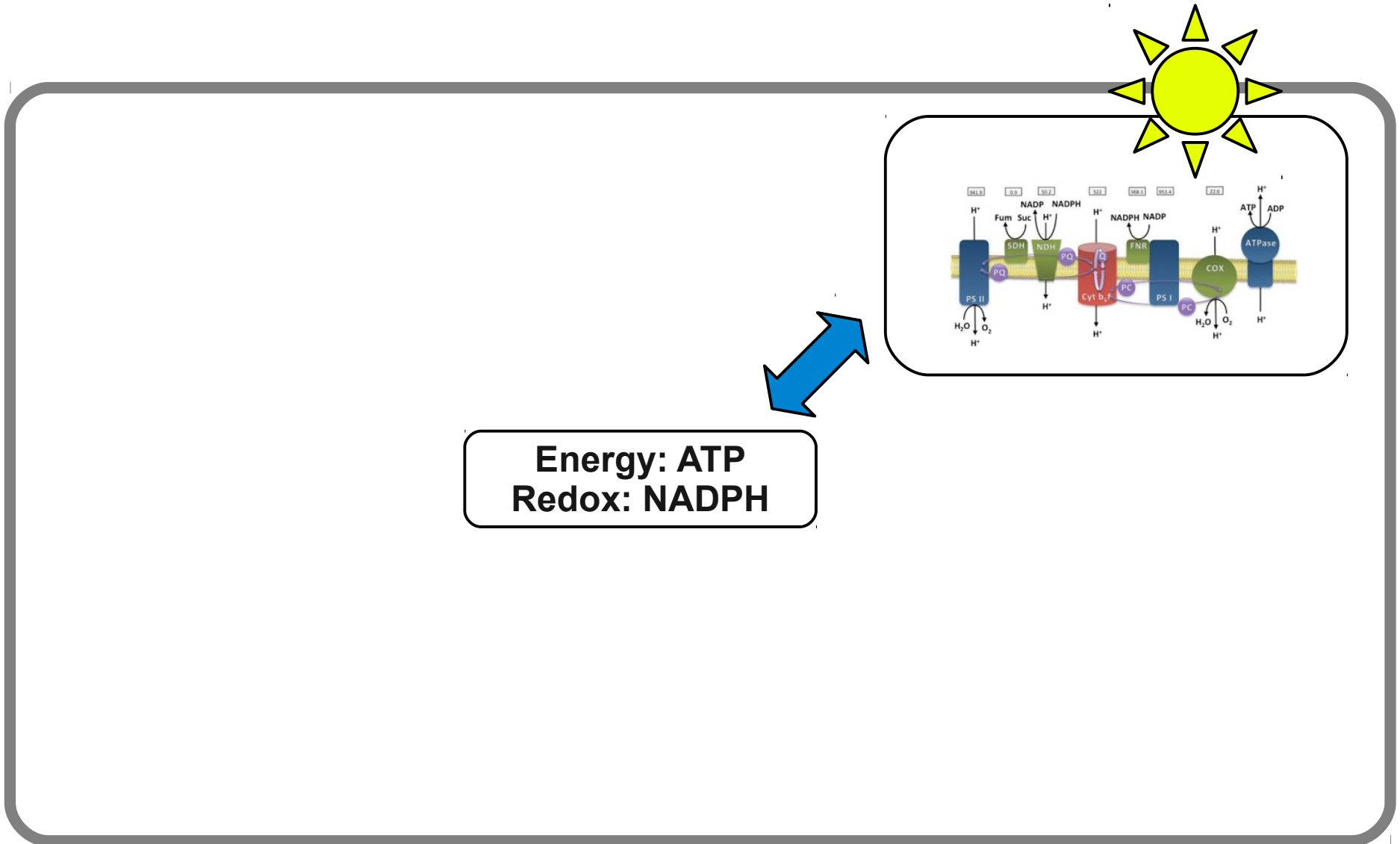


Cyanobacteria: a modular approach

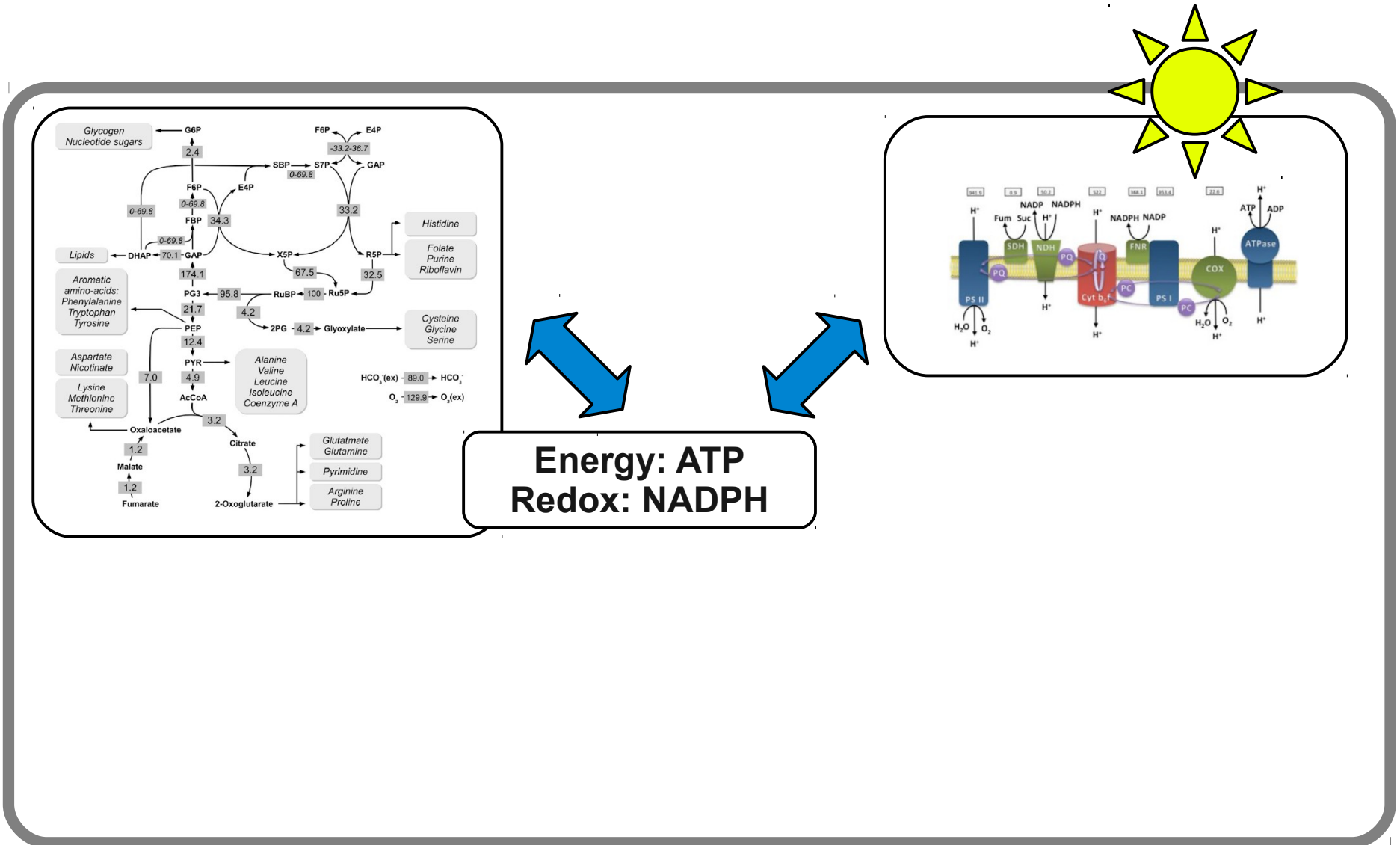
We aim to understand the life and growth of cyanobacteria



Cyanobacteria: a modular approach

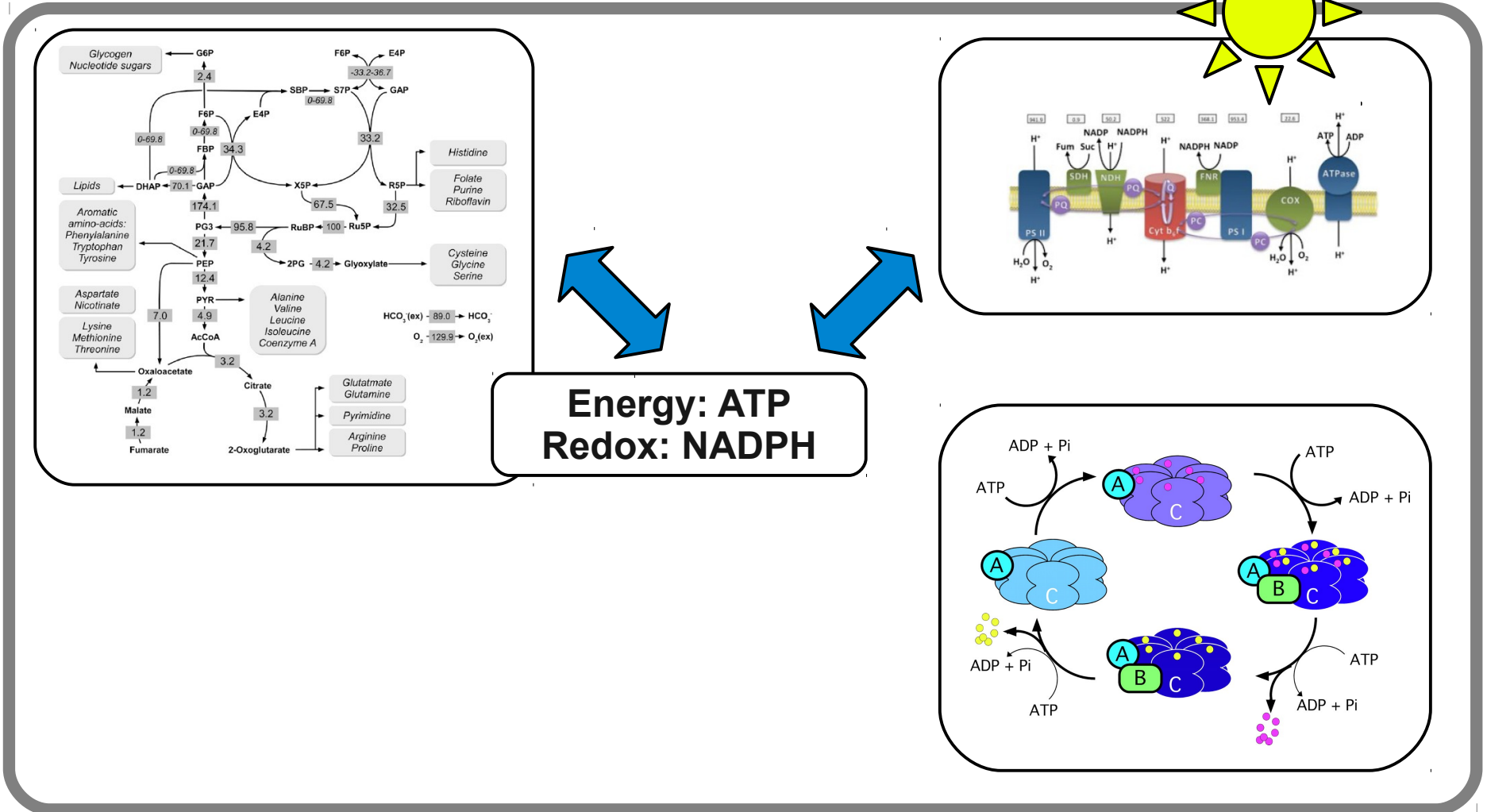


Cyanobacteria: a modular approach



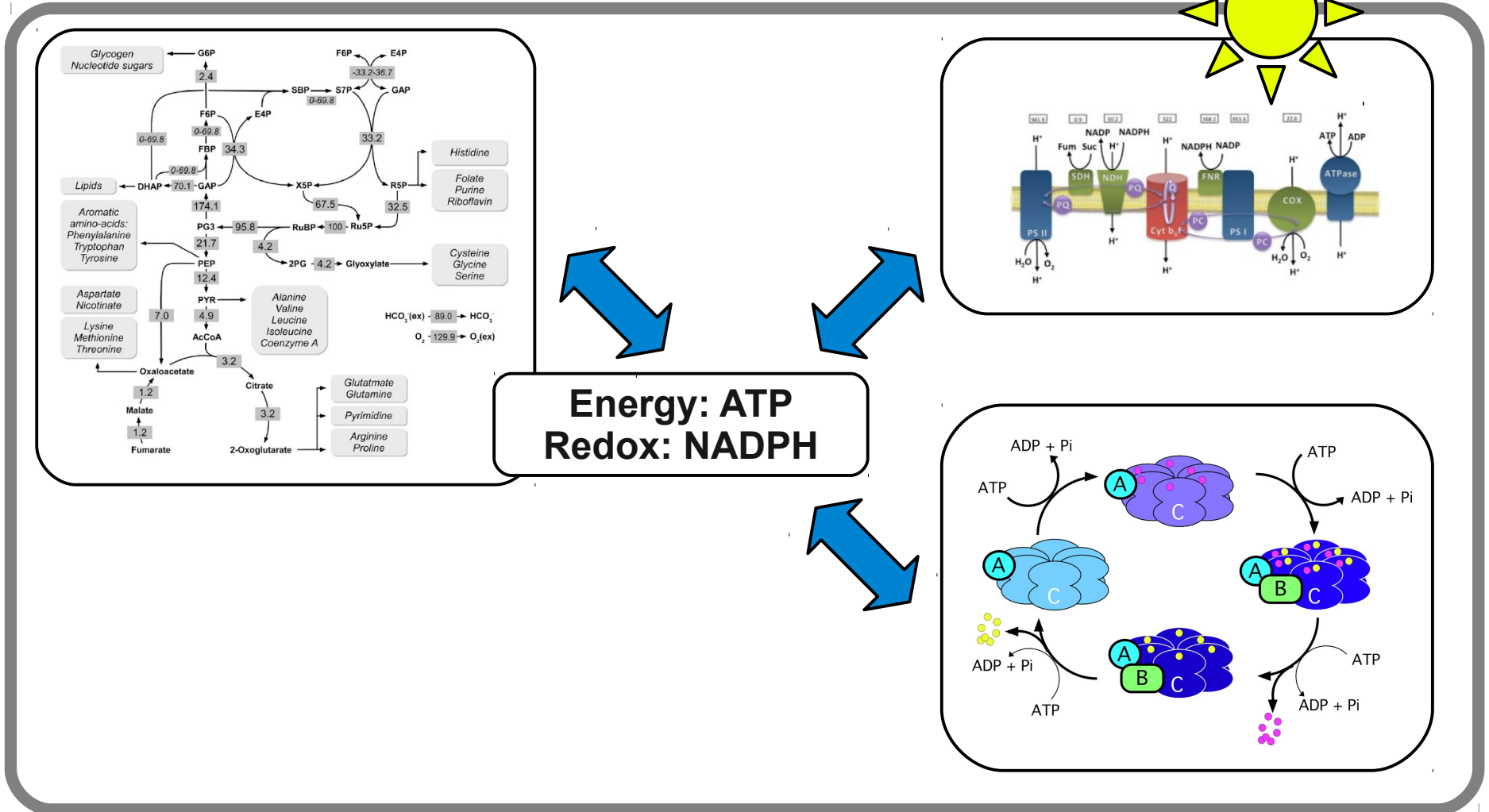


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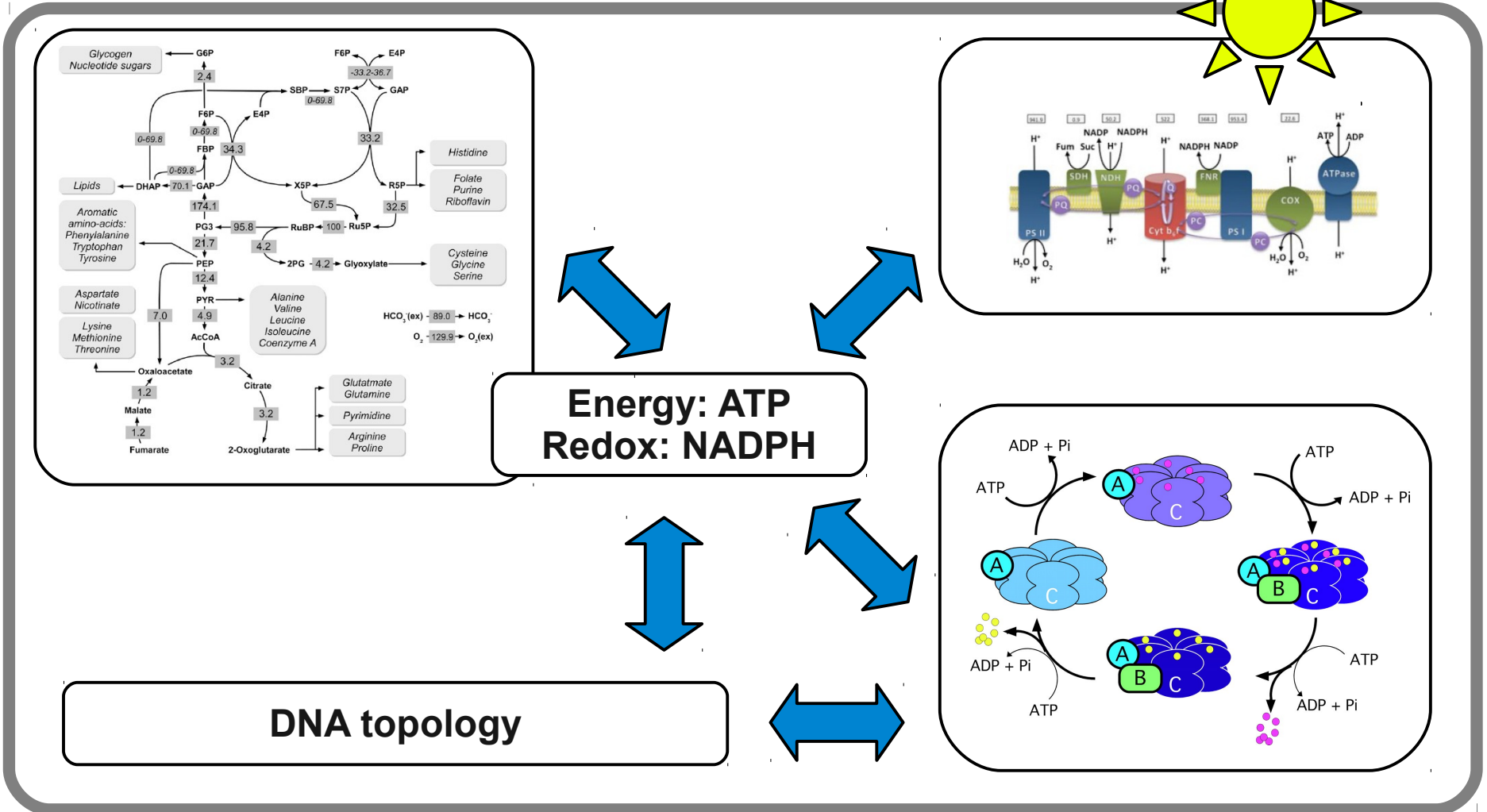


Cyanobacteria: a modular approach



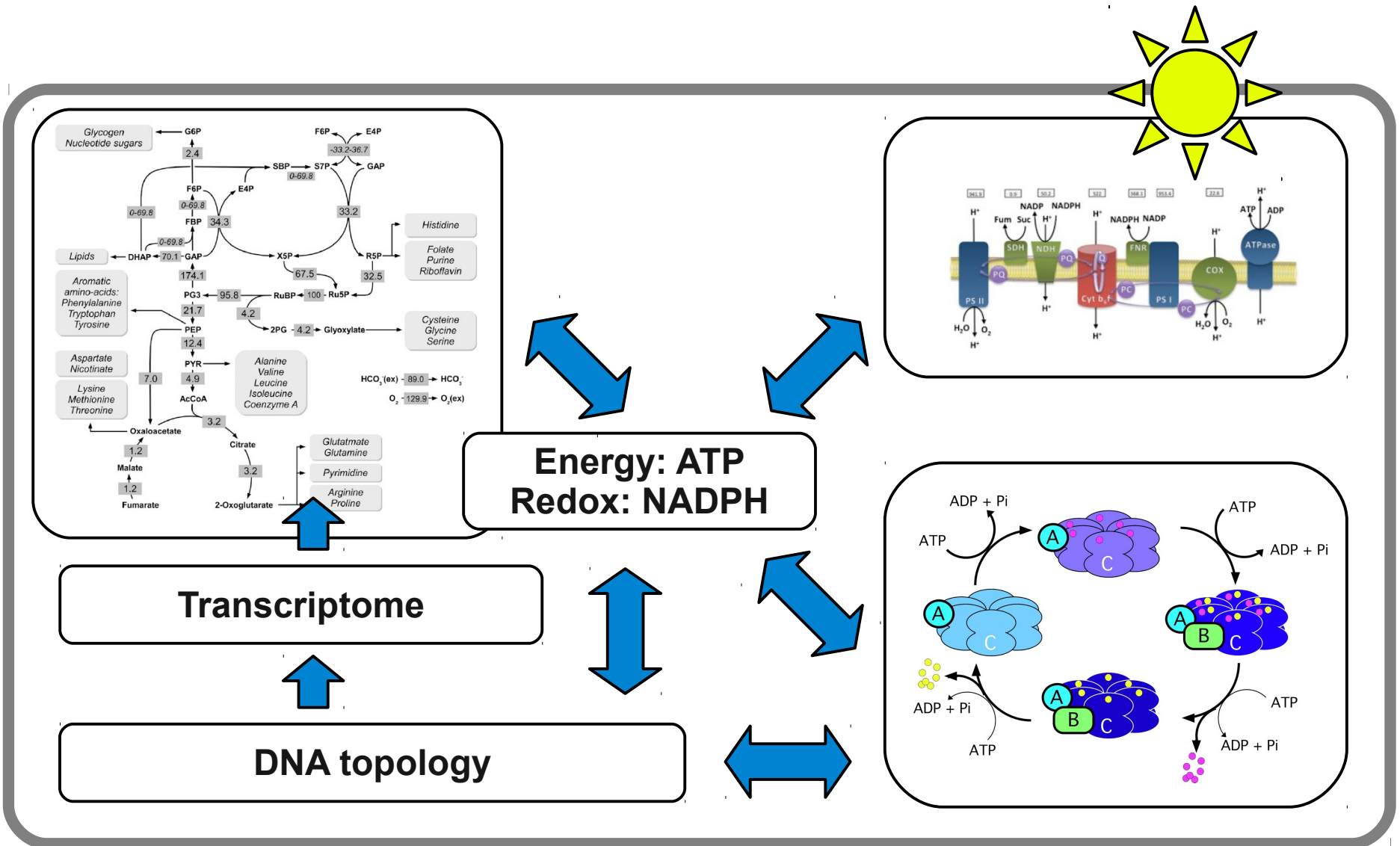


Cyanobacteria: a modular approach



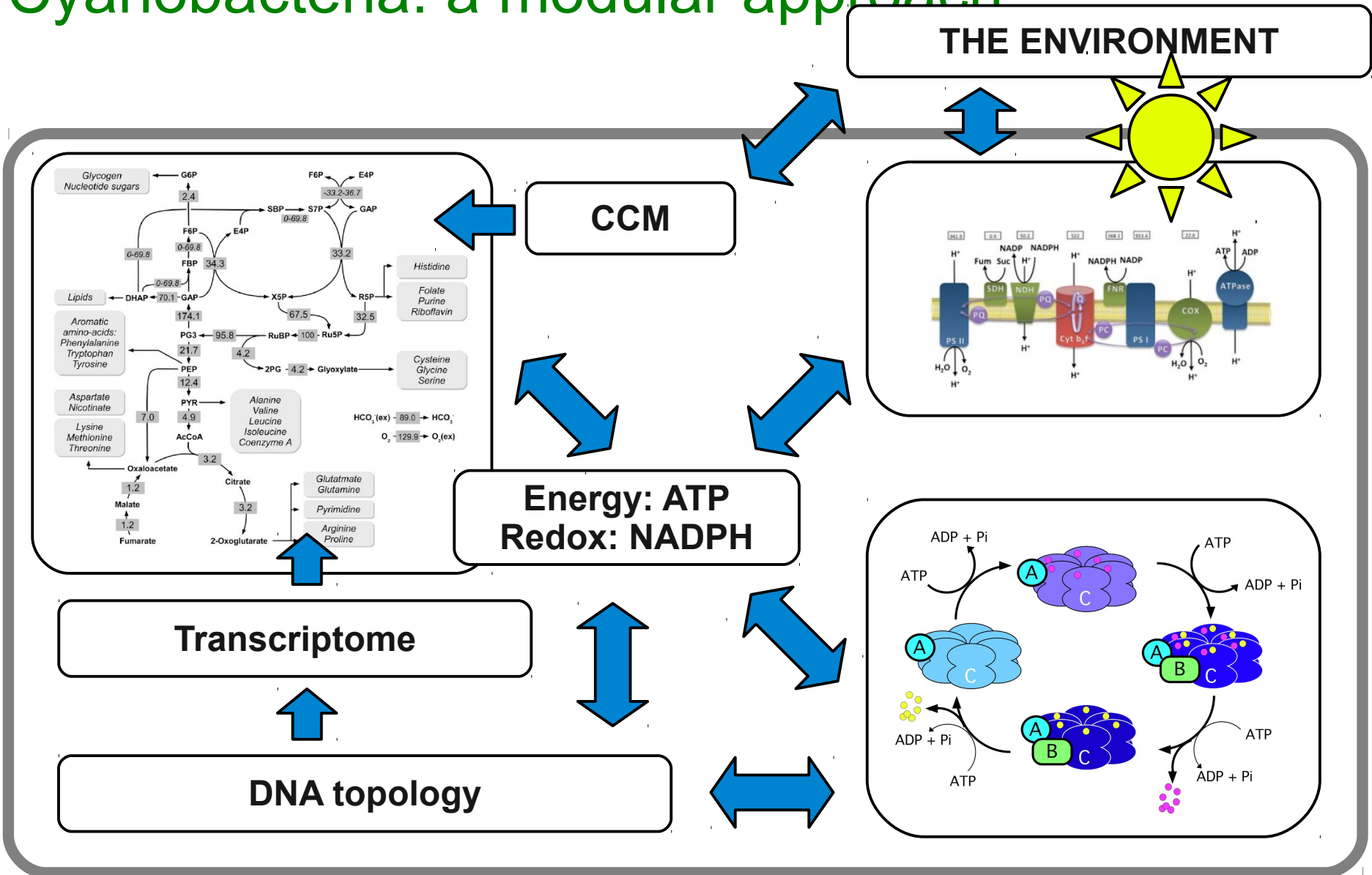


Cyanobacteria: a modular approach





Cyanobacteria: a modular approach





Cyanobacteria: a hierarchy of processes

Phototrophic growth and the environment





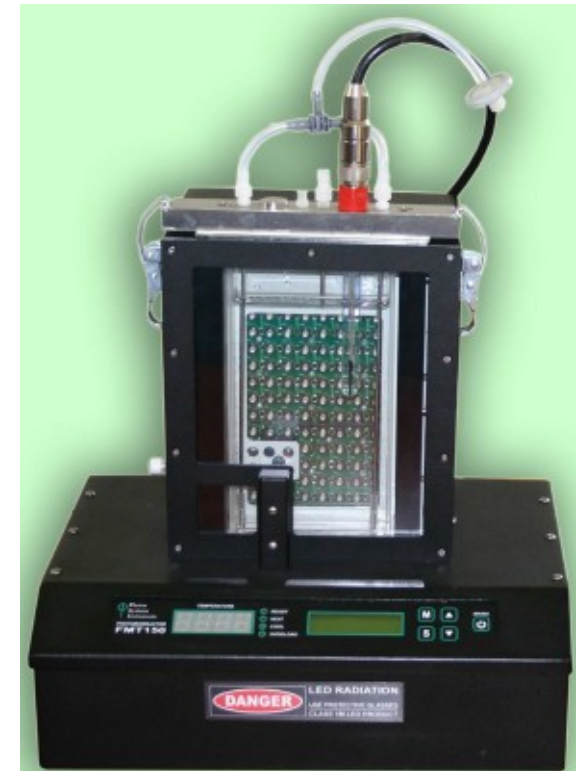
Phototrophic





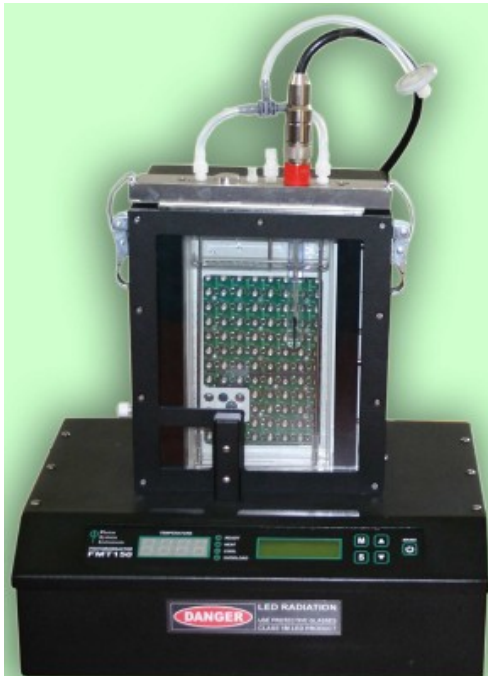
Cyanobacteria: a hierarchy of processes

Phototrophic growth and the environment



Cyanobacteria: a hierarchy of processes

Phototrophic growth and the environment



Traditional ODE model for gas-liquid mass transfer:

$$\frac{dc_l}{dt} = \underbrace{k_{L,b} \frac{A_b}{V_l} (k_H c_b - c_l)}_{\alpha} + \underbrace{k_{L,h} \frac{A_h}{V_l} (k_H c_h - c_l)}_{\beta} + \underbrace{q_{cell}}_{\gamma}$$

$$\frac{dc_b}{dt} = \frac{\Phi}{V_b} (c_{in} - c_b) - \underbrace{k_{L,b} \frac{A_b}{V_b} (k_H c_b - c_l)}_{\alpha}$$

$$\frac{dc_h}{dt} = \frac{\Phi}{V_h} (c_b - c_h) - \underbrace{k_{L,h} \frac{A_h}{V_h} (k_H c_h - c_l)}_{\beta}$$

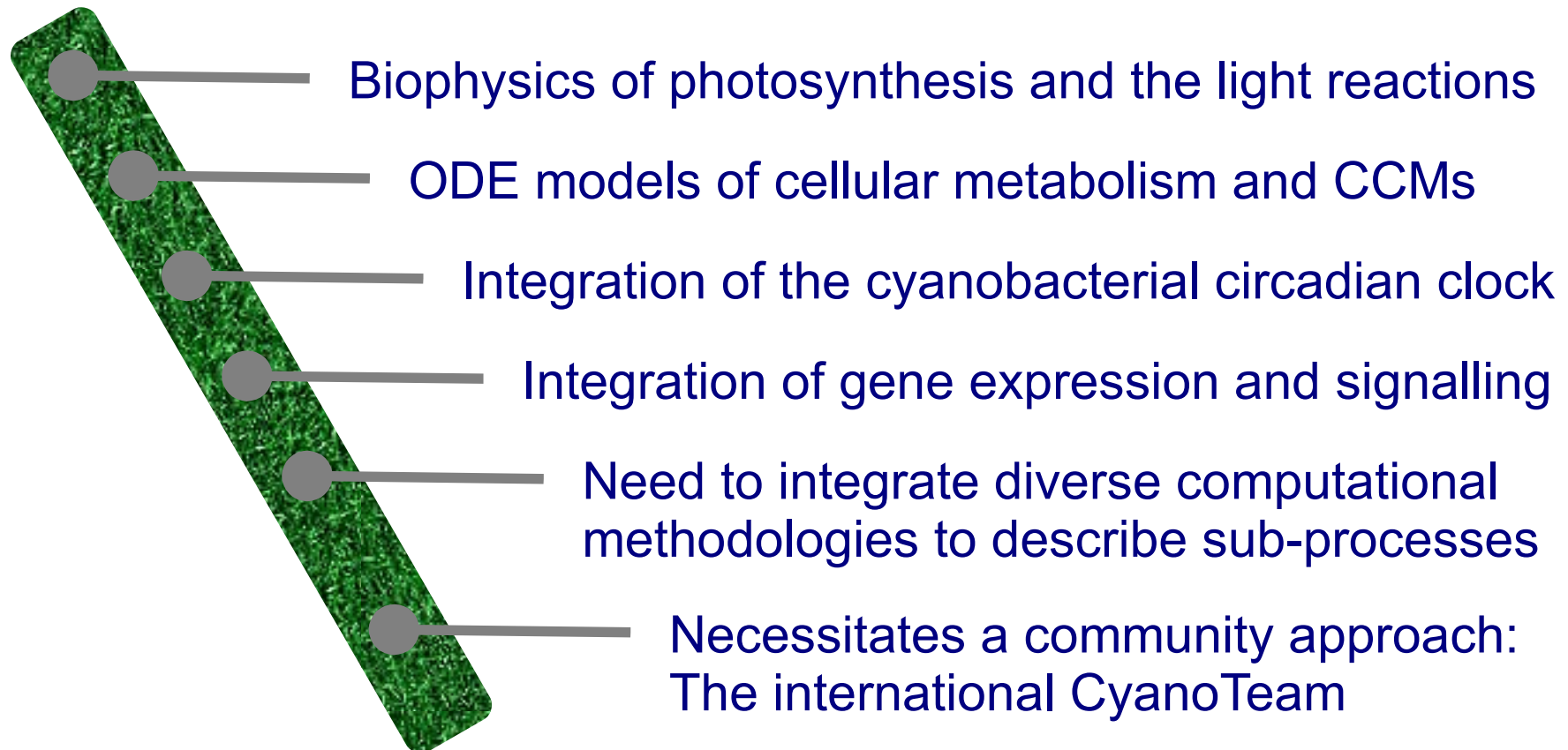
plus carbonate chemistry and a light gradient.

Stefan Mueller et al. An integrated model of photosynthetic growth in a bioreactor: gas-liquid mass transfer, carbonate chemistry, and cellular fluxes (to be completed soon).



Cyanobacteria: a hierarchy of processes

We aim to understand the life and growth of cyanobacteria



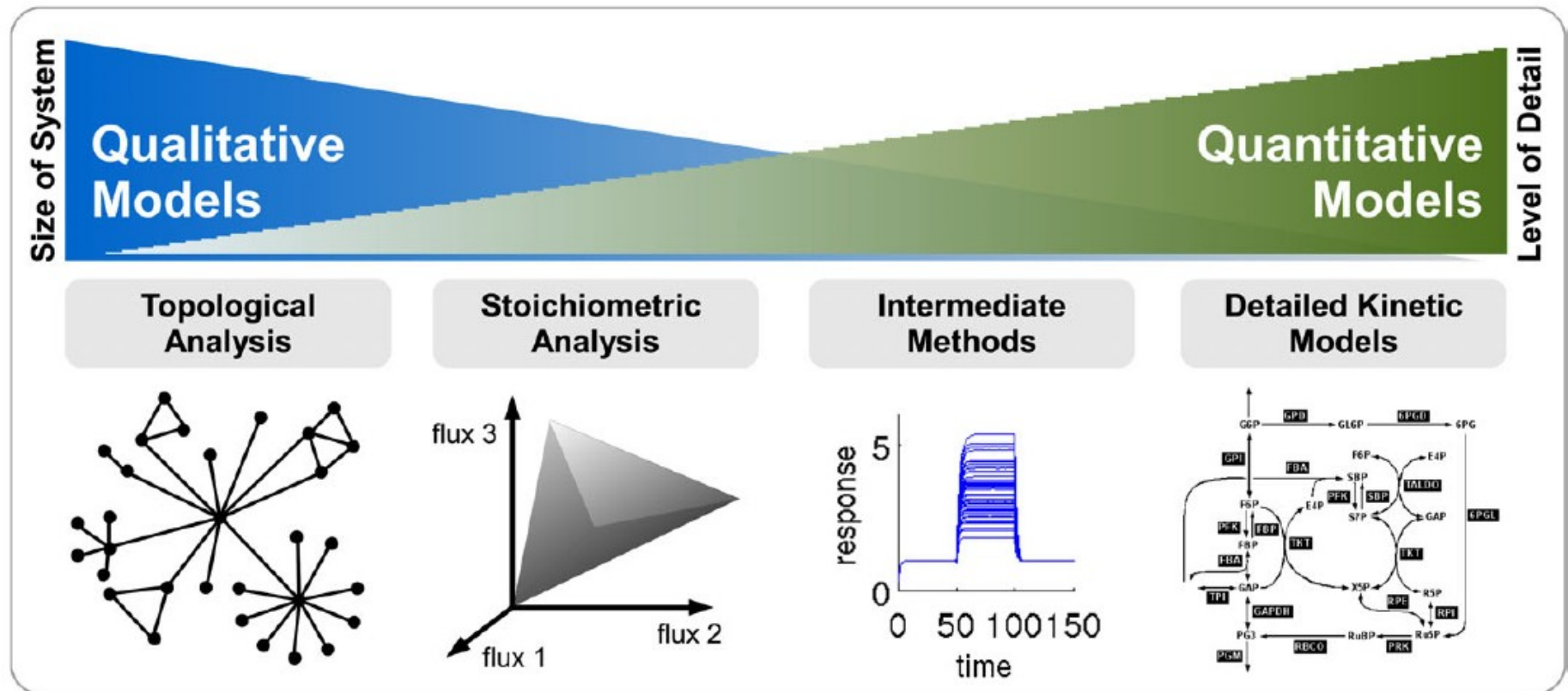


Modelling cellular metabolism

Understanding phototrophic growth in a complex environment

Modelling cellular metabolism

Understanding phototrophic growth in a complex environment





Modelling cellular metabolism

Mechanistic versus teleological models



Based on mechanistic details of the underlying processes (bottom-up)



Modelling cellular metabolism

Mechanistic versus teleological models



Based on constraints and optimization principles (top-down).

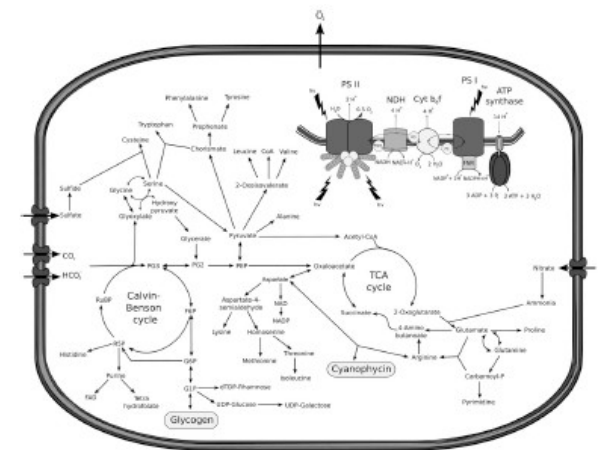
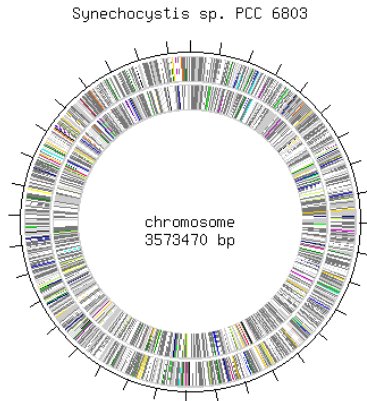
Widely applied to study flux distributions in metabolic network

Modelling cellular metabolism

Mechanistic versus teleological models



All results are based on a high-quality reconstruction of the underlying network of biochemical interconversions.





Modelling cellular metabolism

Mechanistic versus teleological models



All results are based on a high-quality reconstruction of the underlying network of biochemical interconversions.

Metabolic reconstruction: a compendium of all biochemical interconversions of small molecules within a cell.



Modelling cellular metabolism

A stoichiometric model of *Synechocystis* sp. PCC6803



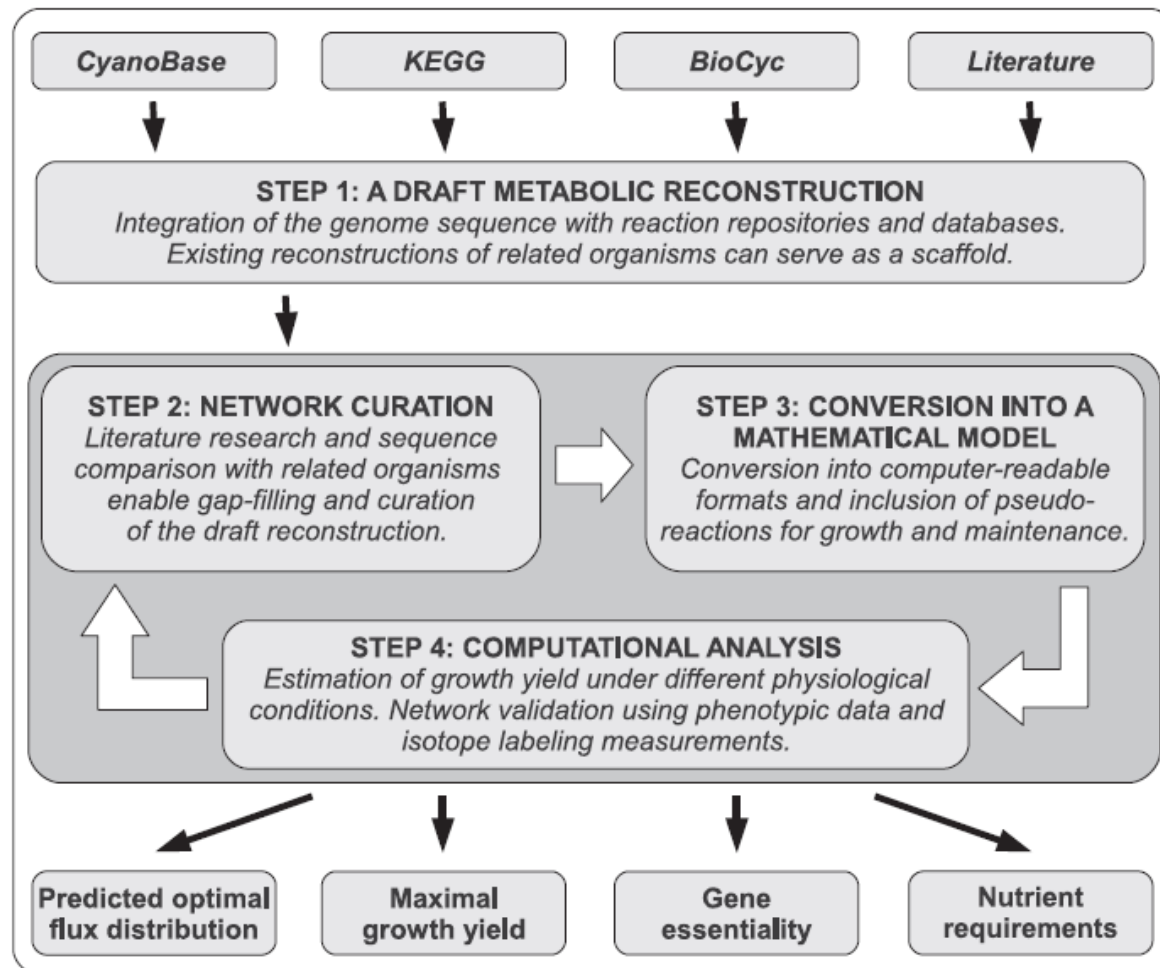
- [1] Start with databases and genome sequence: Initial draft network**
- [2] Identify gaps and inconsistencies: manual curation and literature mining**
- [3] Convert to mathematical model: Include pseudo-reactions for cellular maintenance**
- [4] Analyse the model using constraint-based optimization**

The whole process is iterative and is repeated several times!



Modelling cellular metabolism

A stoichiometric model of *Synechocystis* sp. PCC6803





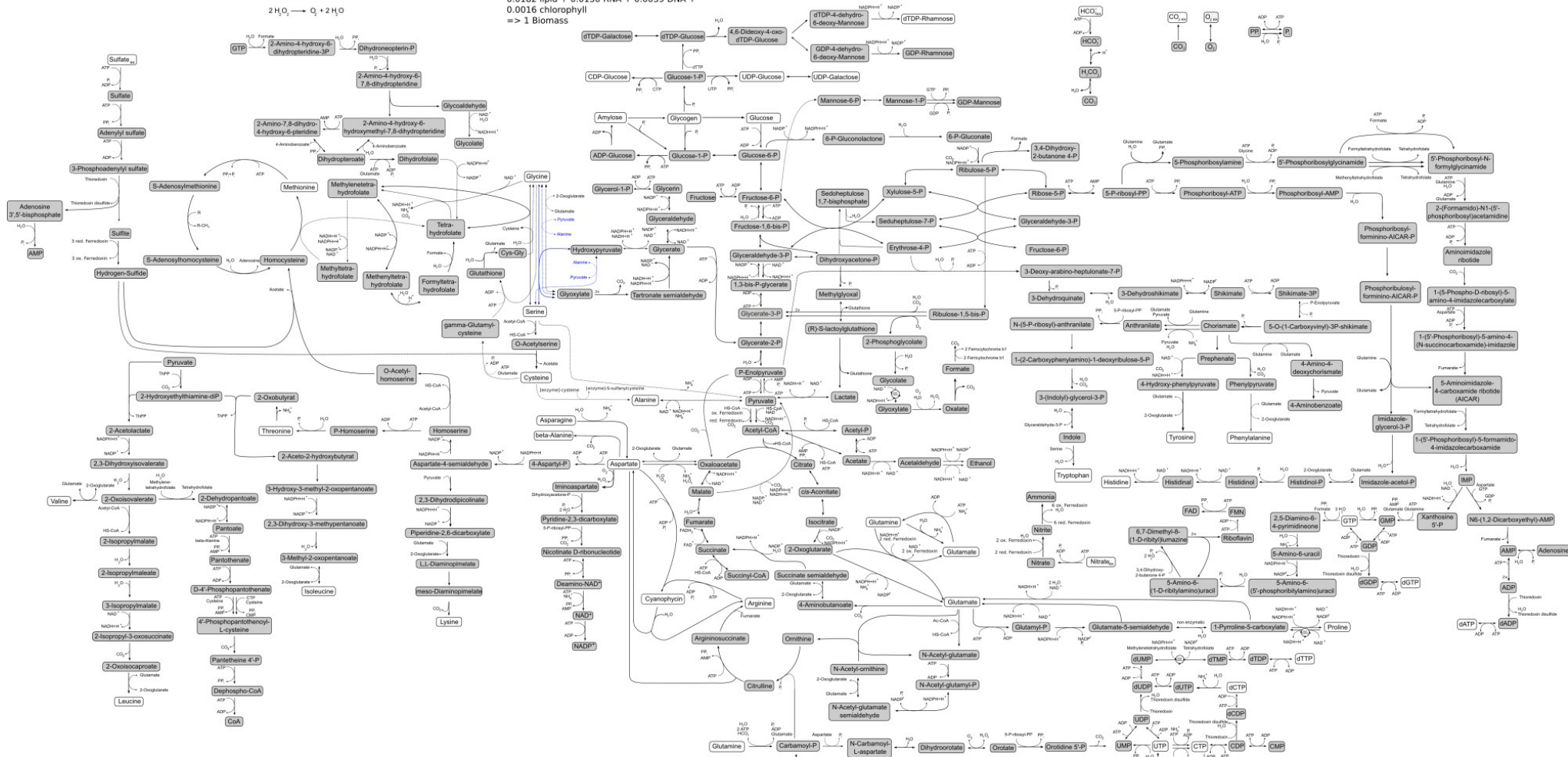
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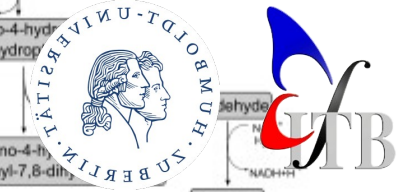
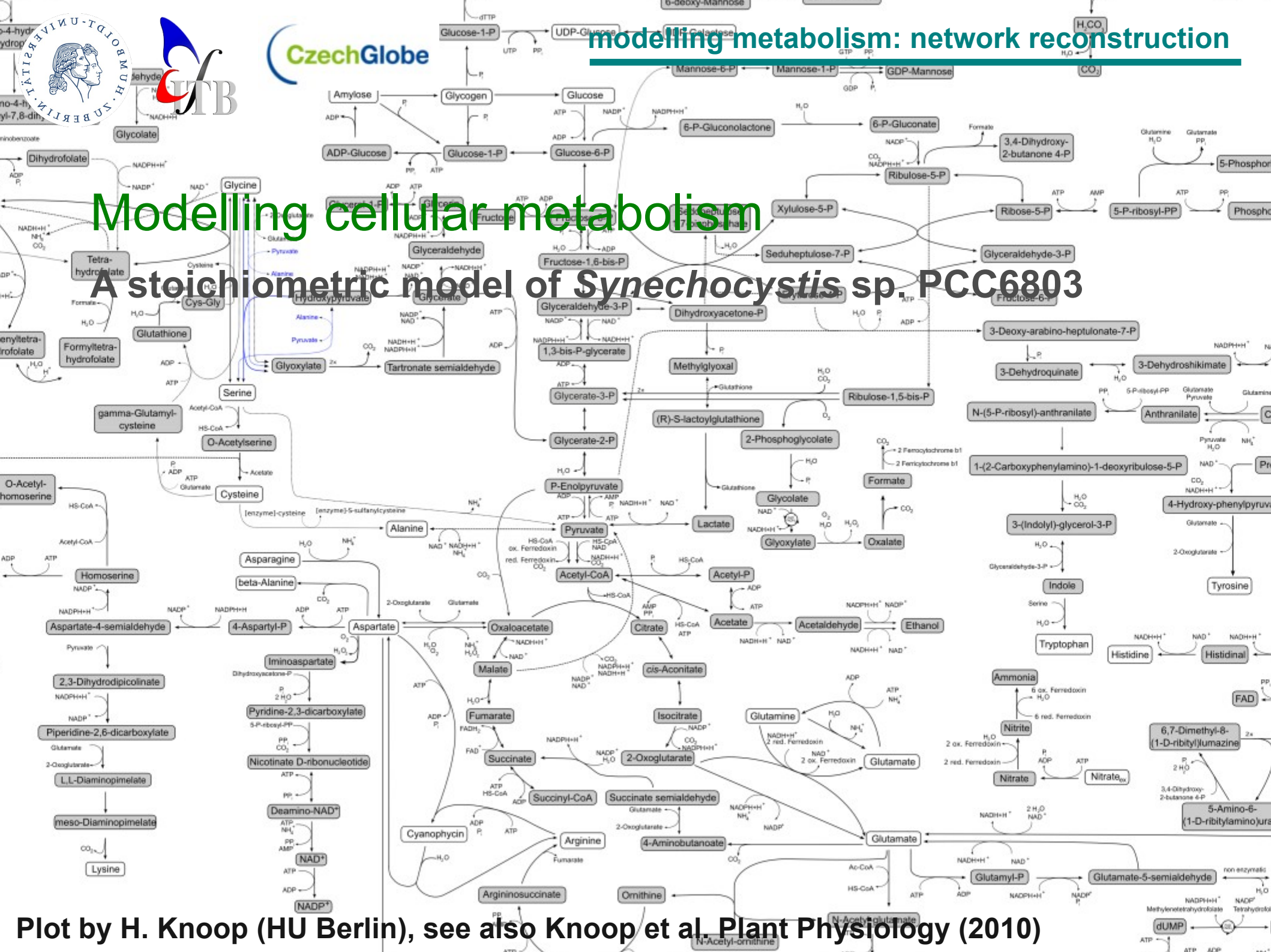
110000000 glucose + 100000000 ATP → 110000000 glucose + 100000000 ATP

Catalase:
 $2\text{H}_2\text{O}_2 \rightarrow \text{O}_2 + 2\text{H}_2\text{O}$

[Bio: 0.8404 protein + 0.1230 carbohydrate +
0.0182 lipid + 0.0130 RNA + 0.0039 DNA +
0.0016 chlorophyll
=> 1 Biomass



Plot by H. Knoop (HU Berlin), see also Knoop et al. Plant Physiology (2010)



Modelling cellular metabolism
 A stoichiometric model of *Synechocystis* sp. PCC6803

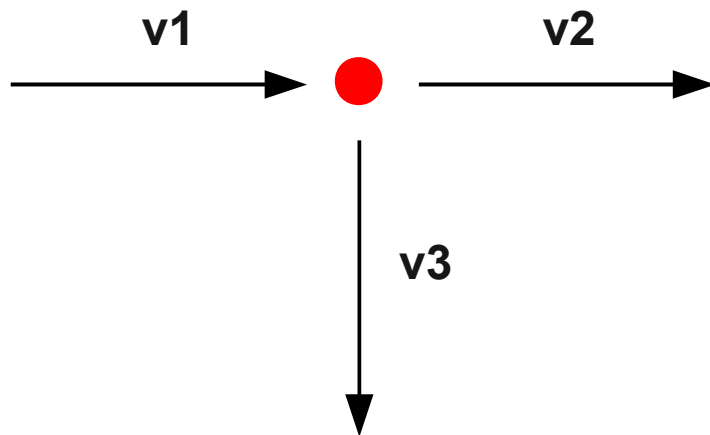


Modelling cellular metabolism

A stoichiometric model of *Synechocystis* sp. PCC6803



Analyse the model using constraint-based optimization



Assuming stationary conditions:

$$v_1 - v_2 - v_3 = 0$$



Modelling cellular metabolism

A stoichiometric model of *Synechocystis* sp. PCC6803



Analyse the model using constraint-based optimization

More general:

$$\frac{dS(t)}{dt} = N\nu(S, k)$$

$$\frac{dS(t)}{dt} = 0 \quad \Rightarrow \quad N\nu(S^0, k) = 0$$

2nd assumptions: metabolic fluxes are organized such that a given (usually linear) objective function Z is maximized.



Modelling cellular metabolism

A stoichiometric model of *Synechocystis* sp. PCC6803



Analyse the model using constraint-based optimization

Constraint-based stoichiometric modeling

$$\text{maximize } Z = w^T \cdot \nu^0$$

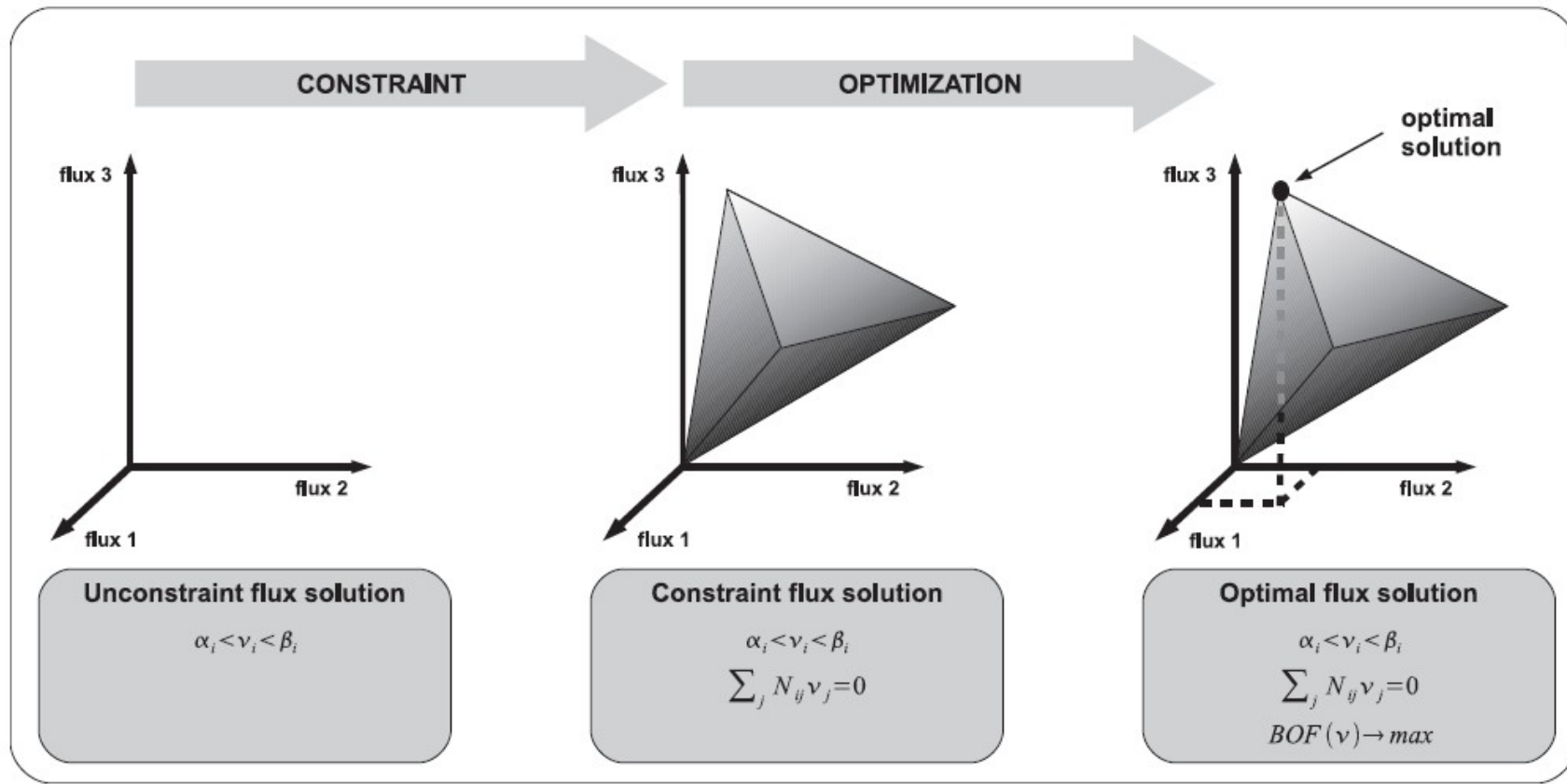
$$\text{subject to: } N\nu^0 = 0$$

$$\text{and } \nu_i^{\min} \leq \nu_i^0 \leq \nu_i^{\max}$$

$$\text{with } i = 1, \dots, r$$

Modelling cellular metabolism

A stoichiometric model of *Synechocystis* sp. PCC6803





A stoichiometric model of *Synechocystis* 6803

Applications of constraint-based optimization

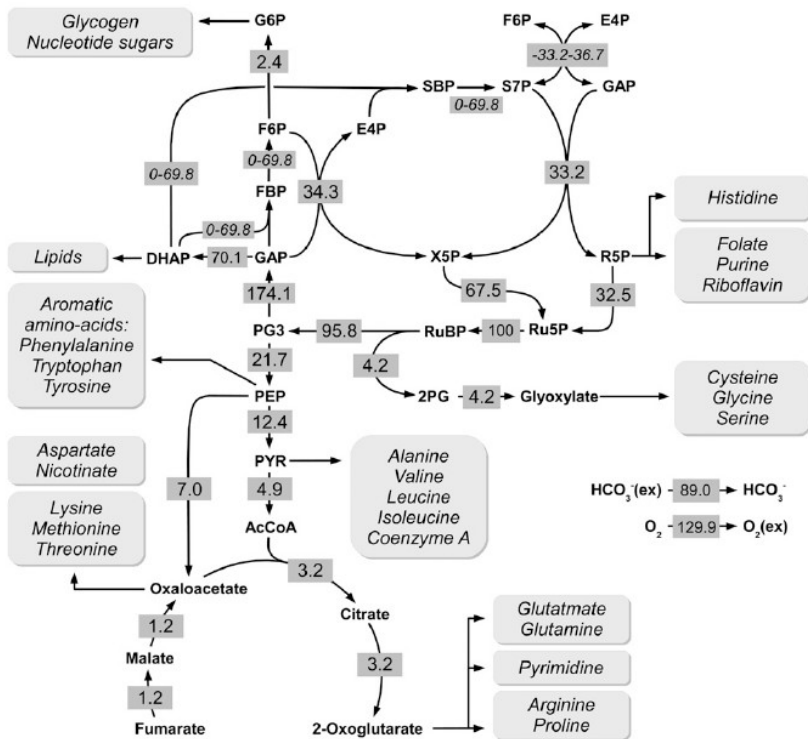
- Optimal flux patterns (maximal biomass yield)
- Flux-variability analysis
- Gene essentiality analysis
- Reaction coupling (with A. Bockmayr, FU Berlin)



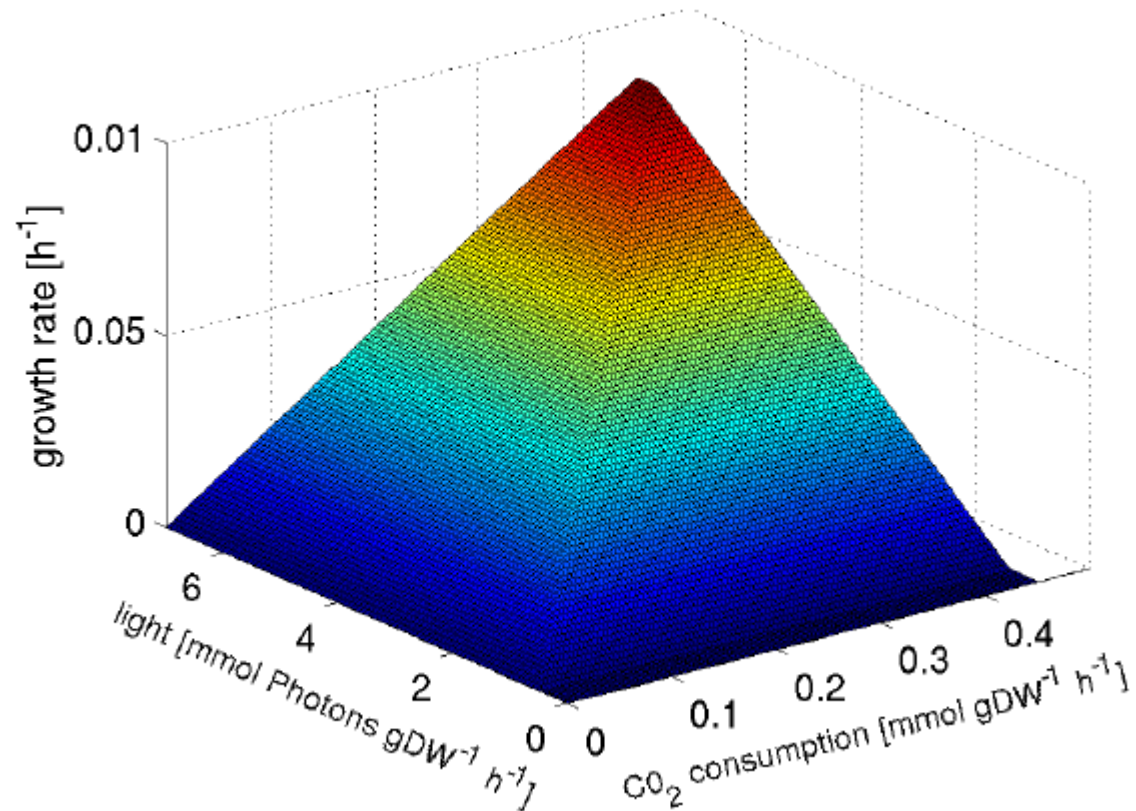
A stoichiometric model of *Synechocystis* 6803

Optimal flux patterns (maximal biomass yield)

flux distribution:



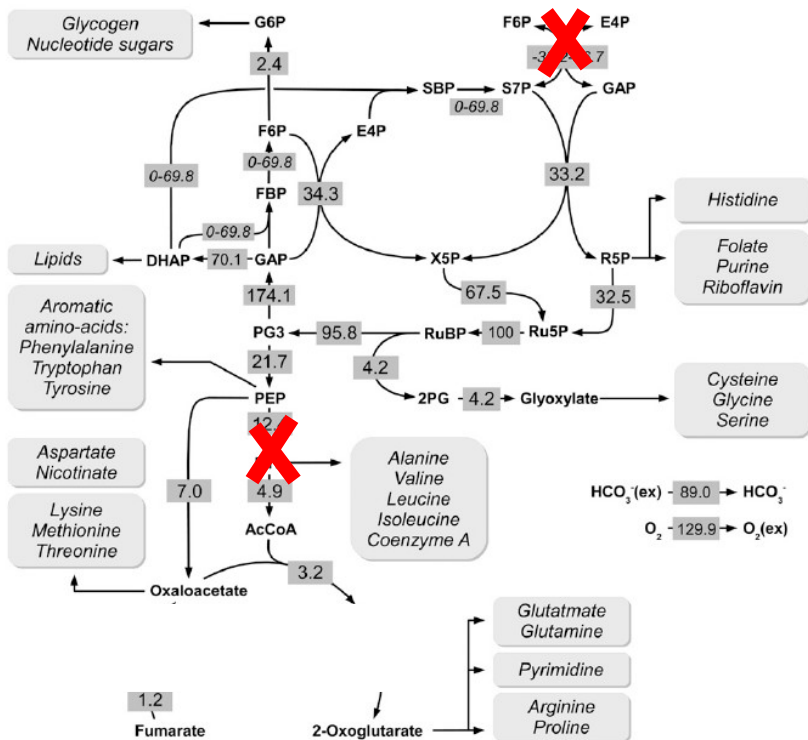
growth rate/yield:





A stoichiometric model of *Synechocystis* 6803

Gene essentiality analysis: network validation



still viable?

126 (of 337) genes are classified as essential for biomass formation: Comparison with CyanoMutants



new hypotheses/questions!



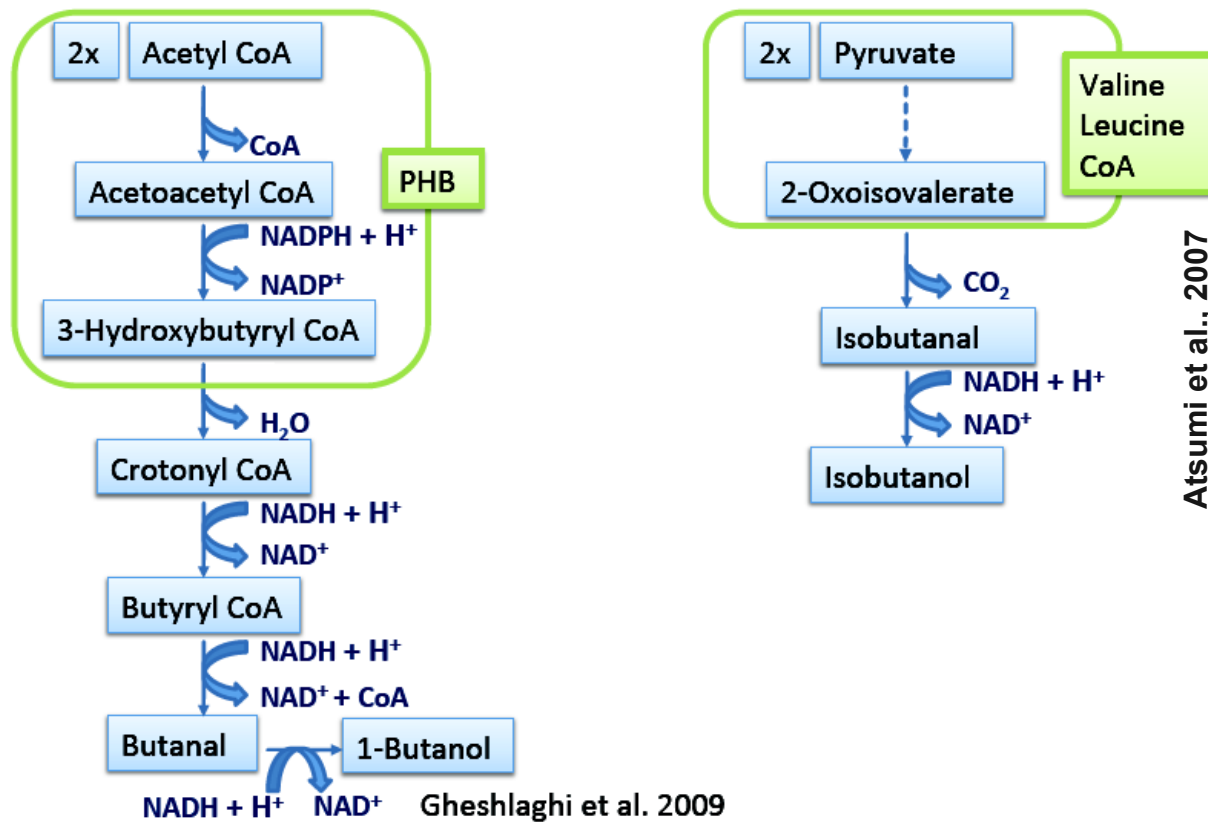
A stoichiometric model of *Synechocystis* 6803

Applications of constraint-based optimization: Biofuels



A stoichiometric model of *Synechocystis* 6803

The model as a platform for strain improvement





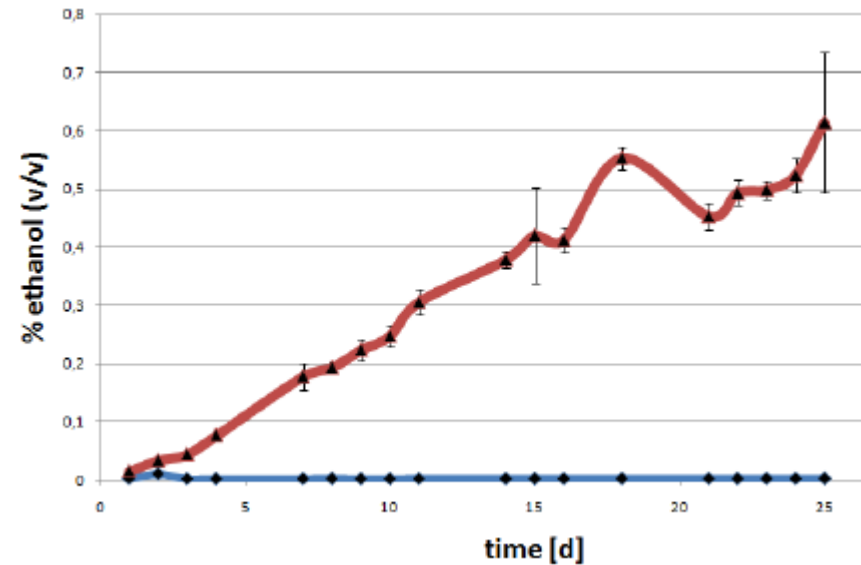
A stoichiometric model of *Synechocystis* 6803

The model as a platform for strain improvement

product	CO ₂	ATP	NAD(P)H	“photons”
ethanol	2	8	6	24.33
ethylene	2.5	23.5	12	64.92
isobutanol	4	18	12	51
isoprene	5	22	13	60.66



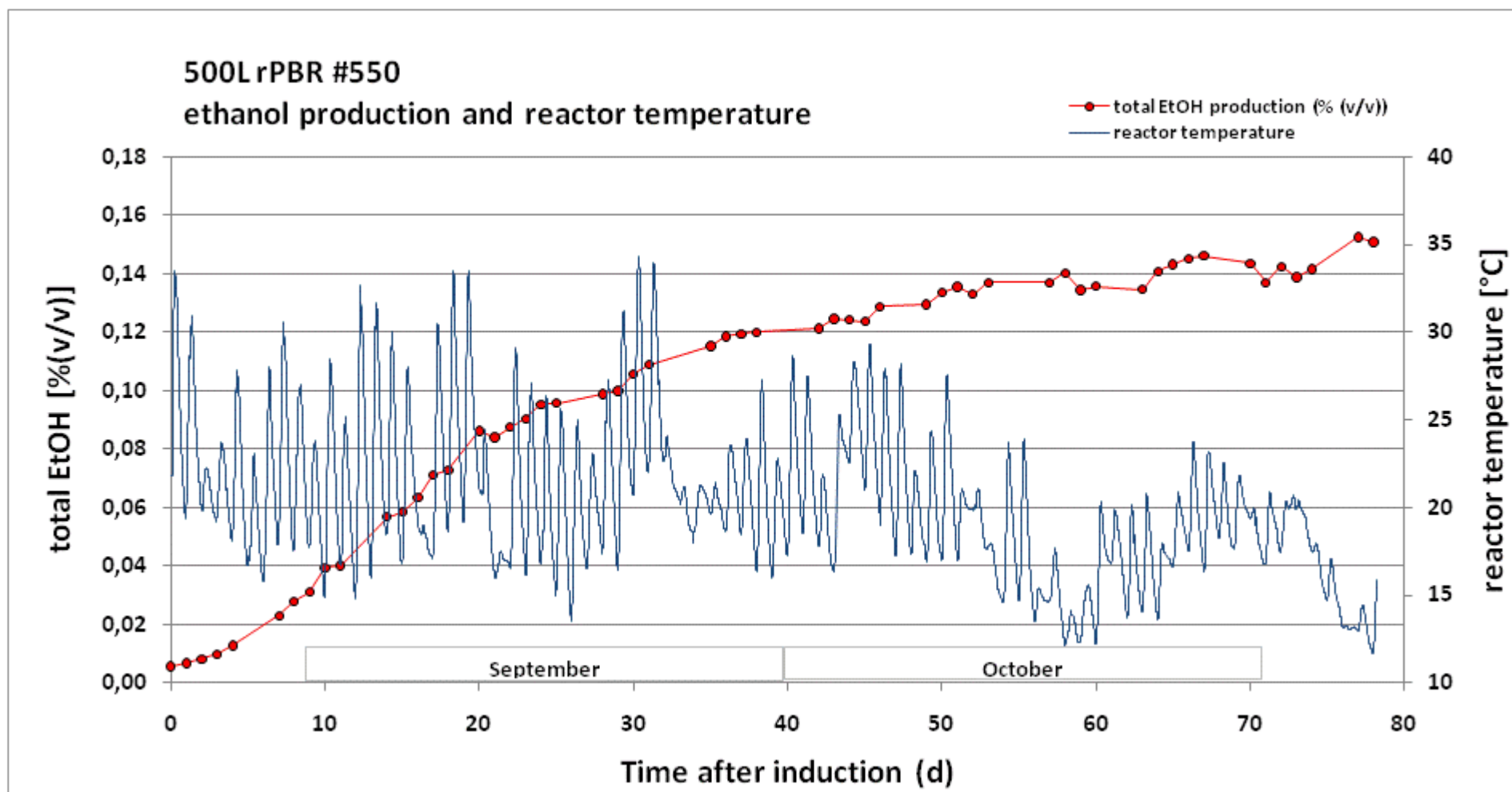
From lab to applications: large-scale cultivation of *Synechocystis* sp. PCC 6803





From lab to applications: large-scale cultivation of *Synechocystis* sp. PCC 6803

Culture Duration: 79 days/Final EtOH Conc. : 0.15 % (v/v)





A stoichiometric model of *Synechocystis* 6803

Applications of constraint-based optimization: Biofuels

Introduce fuel pathways into the stoichiometric reconstruction

- contributions to host optimization and metabolic streamlining
- identify main routes of synthesis for precursor metabolites
- prediction of optimal knockout targets for product formation

**Direct
Fuel**

Direct biological conversion of solar energy to volatile hydrocarbon fuels by engineered cyanobacteria

An EU FP7 collaborative project (grant no. 256808)





A stoichiometric model of *Synechocystis* 6803

CHALLENGES AND EXTENSIONS OF FBA

- Thermodynamic consistency
- The costs of pathways: minimum-cost flow problems
- Temporal coordination of metabolism



A stoichiometric model of *Synechocystis* 6803

Temporal coordination of metabolism



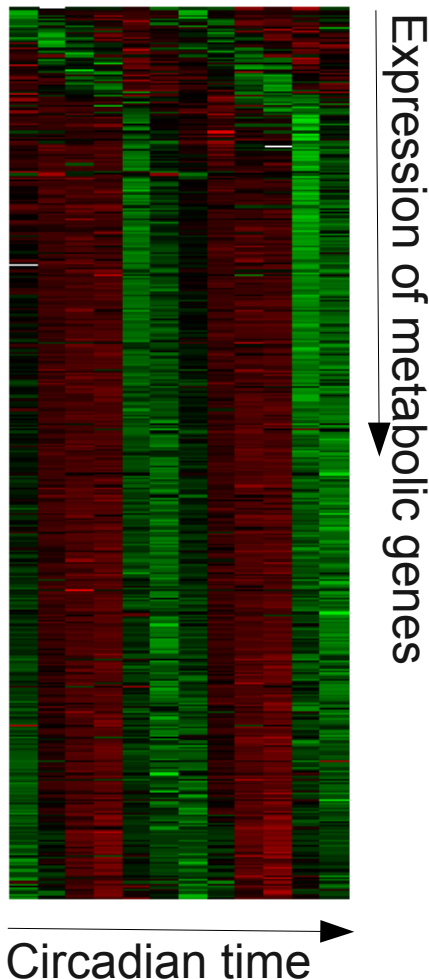
A stoichiometric model of *Synechocystis* 6803

Temporal coordination of metabolism



A stoichiometric model of *Synechocystis* 6803

Temporal coordination of metabolism



Indeed, cyanobacterial metabolism follows a complex circadian program

Most genes expressed during light period

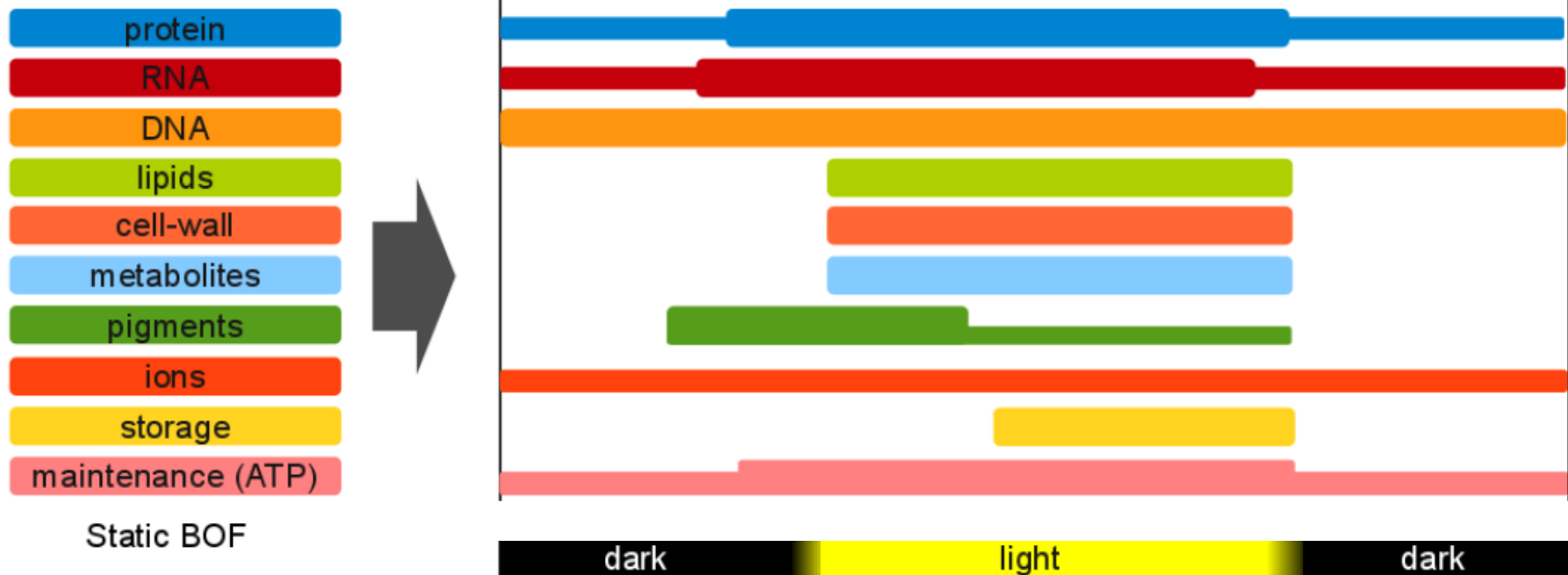
Data: group of I. Axmann (ITB, Berlin)
Clustering/Data analysis: Rob Lehmann, Rainer Machne
submitted



A stoichiometric model of *Synechocystis* 6803

Temporal coordination of metabolism

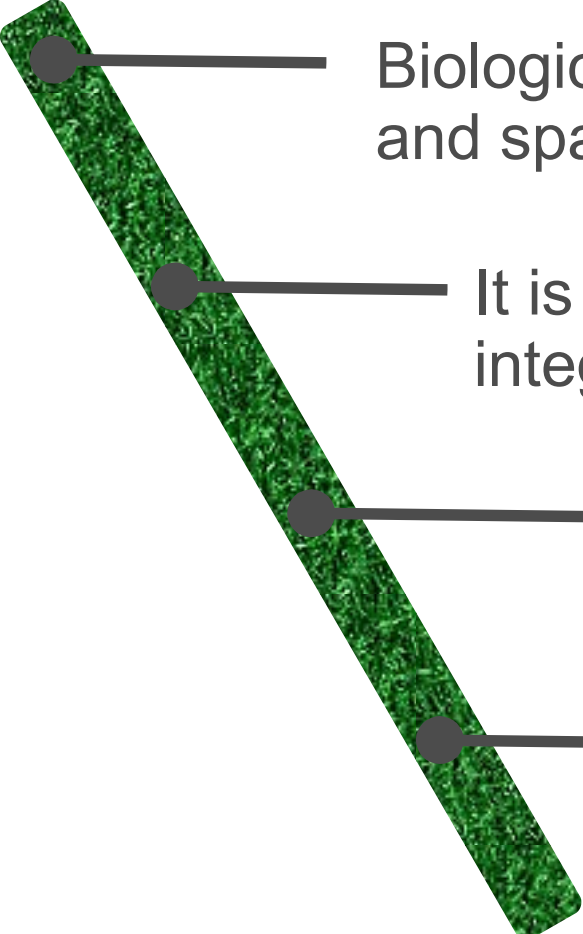
A time-dependent objective function:





Modelling cellular metabolism: summary

Understanding phototrophic growth in a complex environment

- 
- Biological systems typically involve multiple temporal and spatial scales: need for different methodologies.
 - It is a conceptual and computational challenge to integrate diverse systems into a coherent whole.
 - Of most interest are intermediate methods that allow to deal with incomplete and uncertain data.
 - Large-scale predictive models of cells are possible: computational biology needs to integrate parts into a coherent whole



The end

Thanks for your attention!



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