

# Análise sistêmica e engenharia do metabolismo microbiano Biologia Sintética e Biologia de Sistemas.

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José Gregório Cabrera Gomez

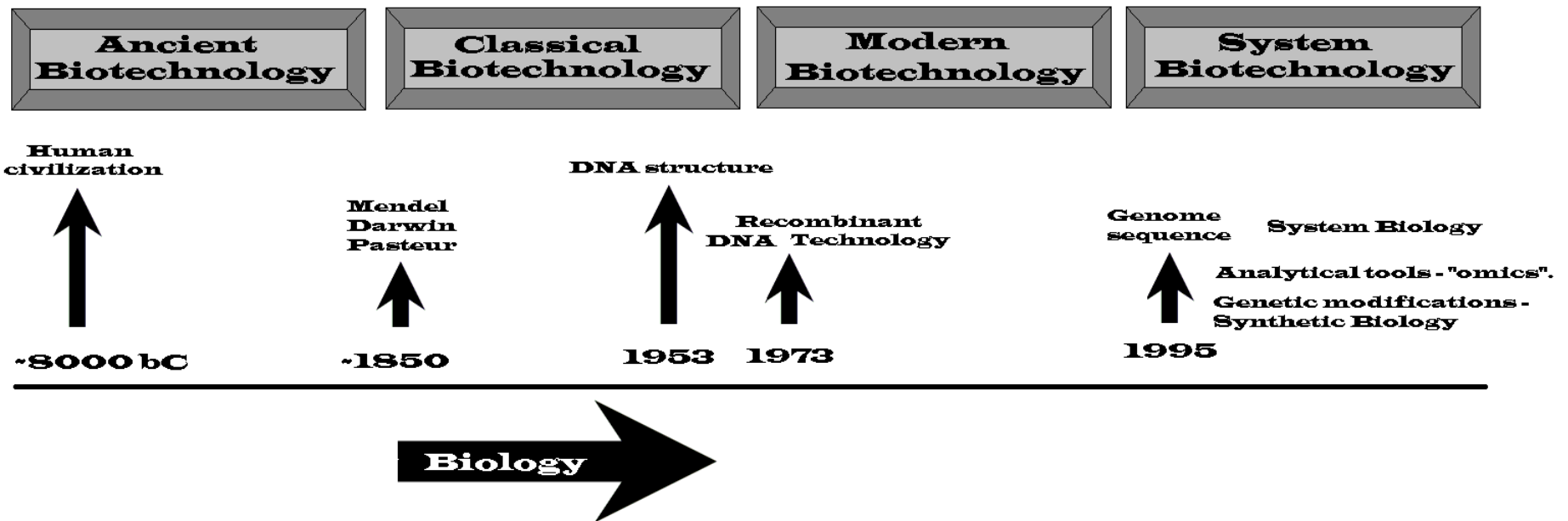


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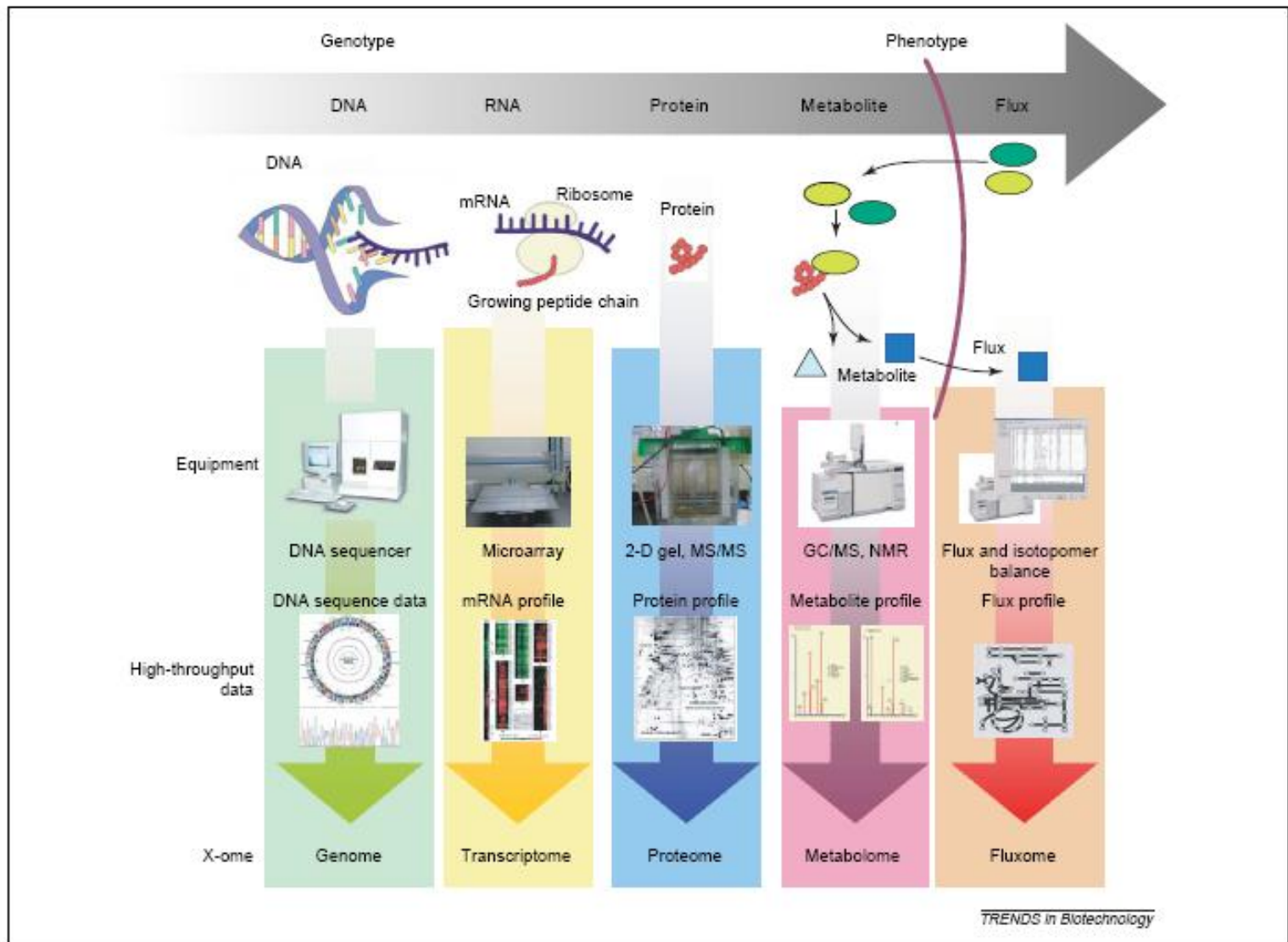
**MICroBiologia**

UNIVERSIDADE DE SÃO PAULO

# Biology and Biotechnology



# ômicas

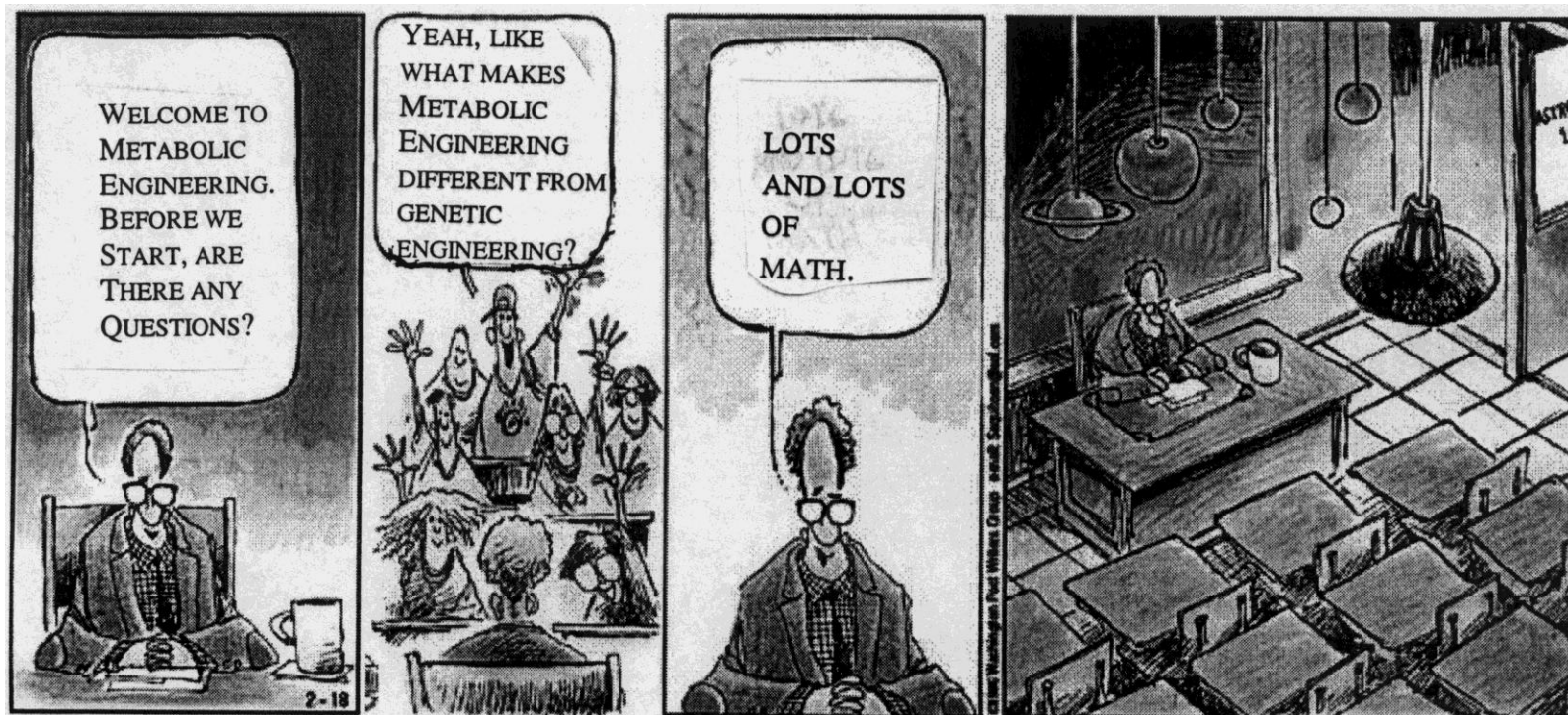


**Figure 1.** High-throughput omics research. Genomics advanced by the development of high-speed DNA sequencing is now accompanied by transcriptome profiling using DNA microarrays. Proteome profiling is joining the high-throughput race as 2D-gel electrophoresis combined with mass spectrometry is advancing. Metabolome profiling is also rapidly advancing with the development of better GC/MS, LC/MS and NMR technologies. Isotopomer profiling followed by challenging with isotopically labeled substrate allows determination of flux profiles in the cell (fluxome).

Metabolic engineering is the improvement of cellular activities by manipulation of enzymatic, transport, and regulatory functions of the cell with the use of recombinant DNA technology. The opportunity to introduce heterologous genes and regulatory elements distinguishes metabolic engineering from traditional genetic approaches to improve strains.

... An interactive cycle of a genetic change, an analysis of the consequences, and the design of a further change...

Toward a Science of Metabolic Engineering.  
James E. Bailey Science, 252: 1668-1675.



# Metabolic Engineering

The knockout or overexpression of genes, usually used in Genetic Engineering, frequently does not result in product yield improvements due a resistance in the metabolism. Therefore, a better knowledge of the metabolism is needed to promote metabolism engineering as a whole to improve biotechnological processes.

**Vallino & Stephanopoulos, 1992**

Metabolic engineering is an enabling science, and distinguishes itself from applied genetic engineering by the use of advanced analytical tools for identification of appropriate targets for genetic modifications and possibly even the use of mathematical models to perform *in silico* design of optimized cell factories.

**Nielsen & Jewett, 2008 FEMS Yeast Res.**

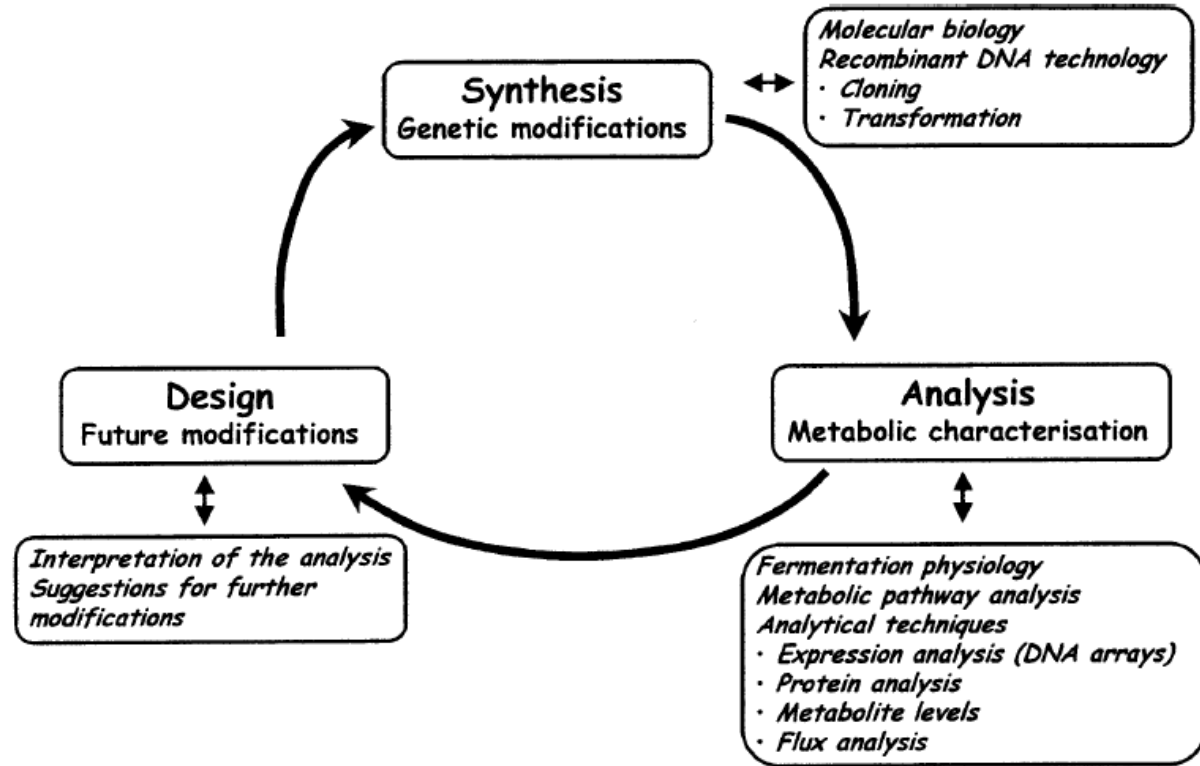


TABLE 1. Overview of reactions, metabolites, and ORFs in reconstructed metabolic networks<sup>a</sup>

Organism	No. of reactions	No. of metabolites	No. of metabolic ORFs	Total no. of ORFs	% of ORFs involved in metabolism
<i>H. pylori</i>	444	340	268	1,638	16
<i>H. influenzae</i>	477	343	362	1,880	19
<i>E. coli</i>	720	436	695	4,485	15
<i>S. cerevisiae</i>	1,175	584	708	5,773	12 <sup>b</sup>

<sup>a</sup> The reconstructed networks are described in references 6, 8, 17, and 18.

<sup>b</sup> The value is based on a recent gene count (3).

Table 3. Frequency of precursor metabolites and cofactors in a *Saccharomyces cerevisiae* genome scale model\*

Precursor metabolite	No of reactions	Cofactor	No of reactions
Glucose-6P	16	ATP	188
Fructose-6P	18	ADP	146
Ribose-5P	20	NADH	65
Erythrose-4P	6	NAD <sup>+</sup>	78
Glyceraldehyde-3P	13	NADPH	78
3-Phosphoglycerate	6	NADP <sup>+</sup>	86
Phosphoenolpyruvate	12		
Pyruvate	27		
Acetyl-CoA	32		
2-Oxoglutarate	38		
Succinyl-CoA	3		
Oxaloacetate	12		

\*The data are taken from the metabolic model developed by Forster *et al.* (2003).

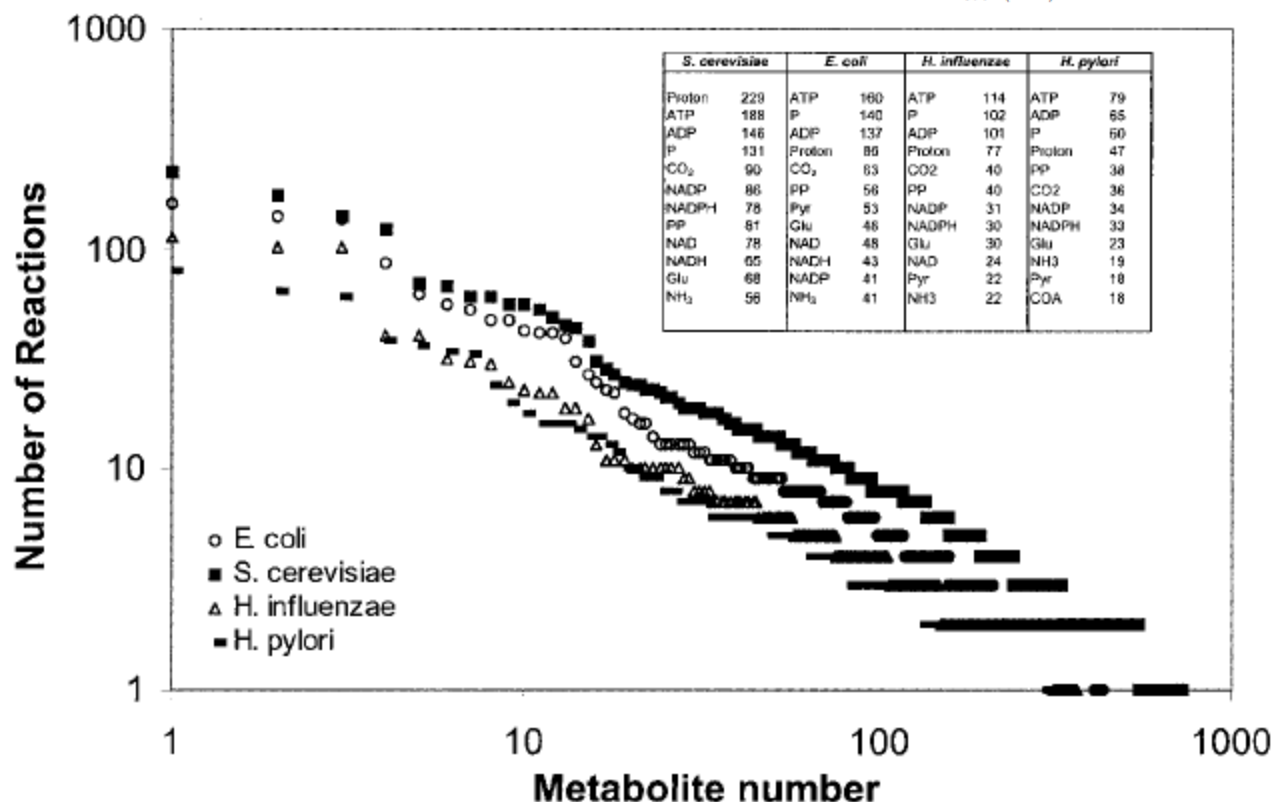
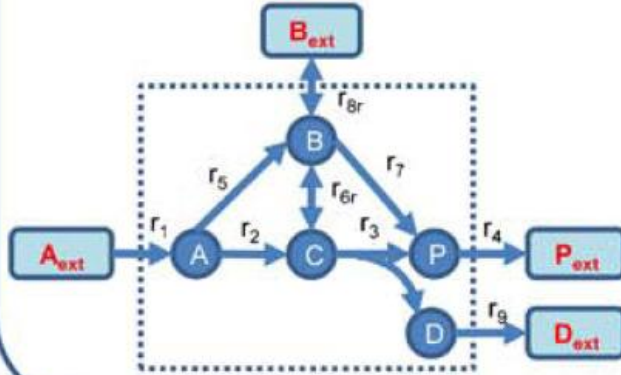


FIG. 1. Frequency plot of the number of reactions that each metabolite appears in for four different reconstructed metabolic networks. For each metabolic network the 10 metabolites that appear in the most reactions are listed. PP, pyrophosphate; COA, coenzyme A. The numbers in the box specify the numbers of reactions the 10 most frequently used metabolites participate in for the four different microorganisms.

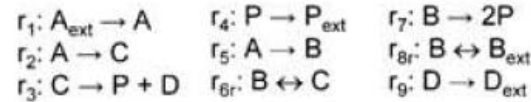
# analysis of cellular metabolism

## Problem statement

### Network



## Stoichiometric reactions



## Equations to solve

A

$$\underline{S} \cdot \underline{r} = \underline{0}$$

Thermodynamic constraints:  
 $r_{1,5,7,9} \geq 0$

## Stoichiometric matrix

	r <sub>1</sub>	r <sub>2</sub>	r <sub>3</sub>	r <sub>4</sub>	r <sub>5</sub>	r <sub>6r</sub>	r <sub>7</sub>	r <sub>8r</sub>	r <sub>9</sub>
A	1	-1	0	0	-1	0	0	0	0
B	0	0	0	0	1	-1	-1	-1	0
C	0	1	-1	0	0	1	0	0	0
D	0	0	1	0	0	0	0	0	-1
P	0	0	1	-1	0	0	2	0	0

$\underline{r} = [r_1 \ r_2 \ r_3 \ r_4 \ r_5 \ r_{6r} \ r_7 \ r_{8r} \ r_9]^T$

$$\frac{d}{dt} \underline{C} = \underline{S} \times \underline{r} - \mu \times \underline{C},$$

$\mu \cdot C$  (negligible)

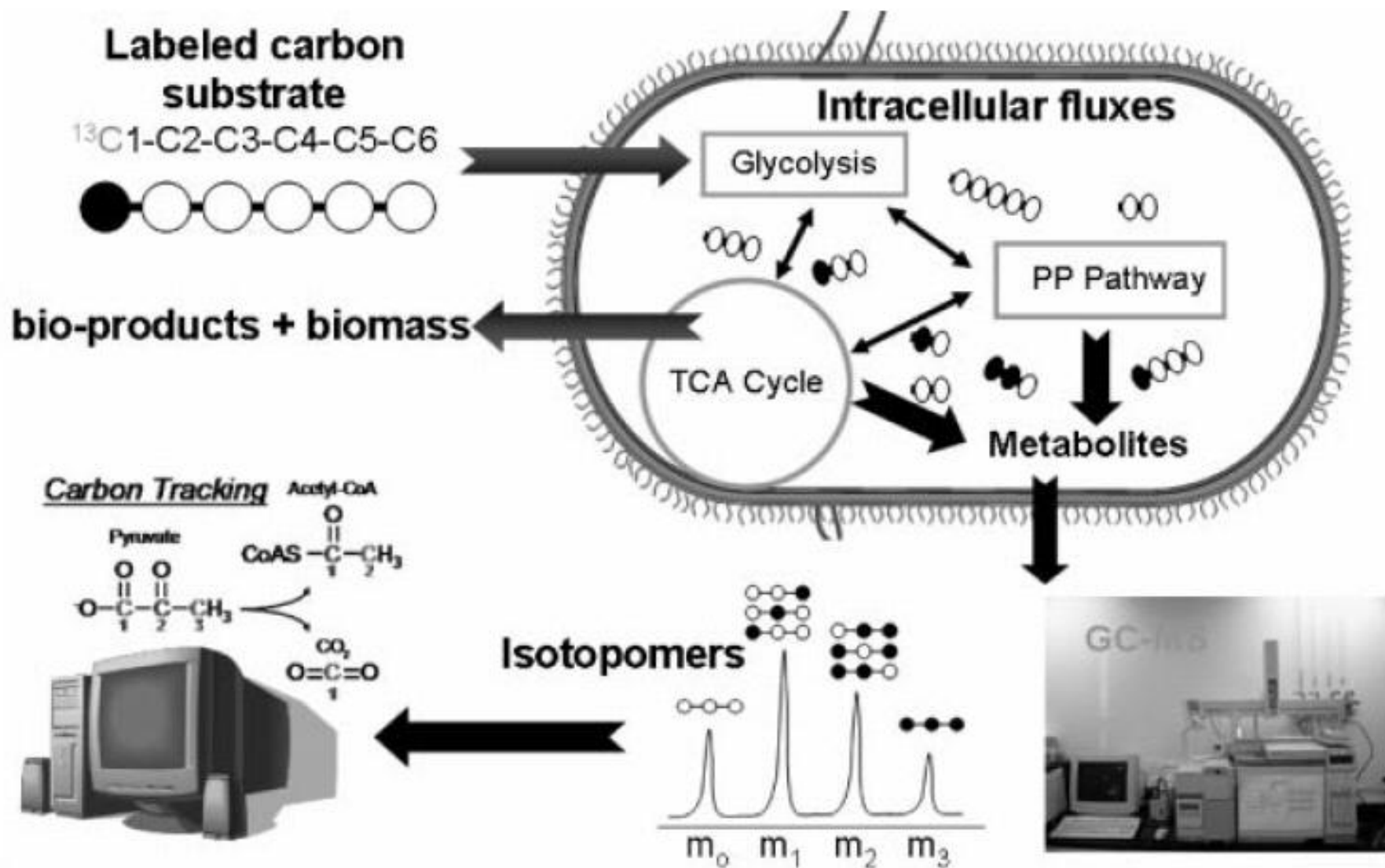
$dC/dt = 0$  (steady state)

$$S \cdot r = 0 \text{ (Eq 2)}$$

$$r_i \geq 0 \text{ (Eq 3)}$$

Tools for analysis of cellular metabolism can be grouped into three categories, all of them developed from the same mathematical model:

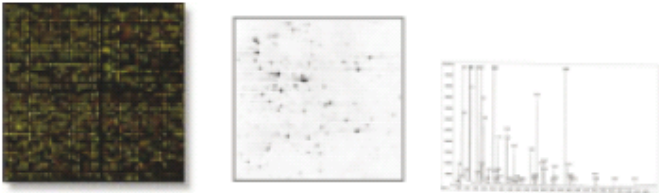
- (1) Metabolic flux analysis,
- (2) Flux balance analysis and
- (3) Metabolic pathway analysis (Elementary mode analysis).



**FIGURE 2.** Protocol for  $^{13}\text{C}$ -based flux analysis. [Color figure can be viewed in the online issue, which is available at [www.interscience.wiley.com](http://www.interscience.wiley.com).]



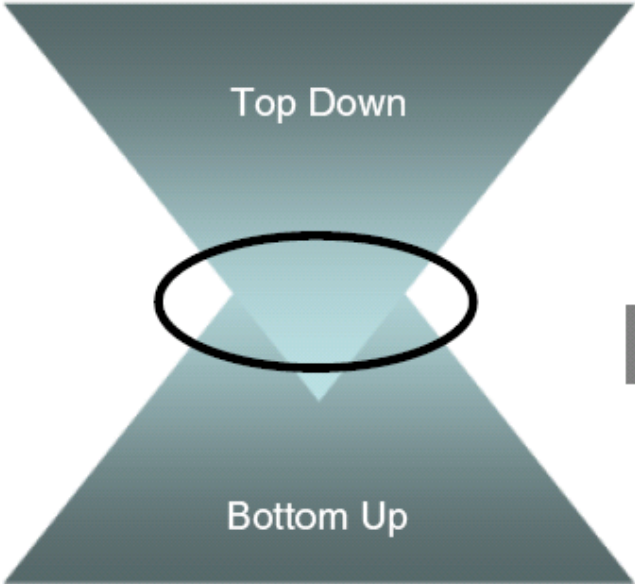
# Metabolic Engineering



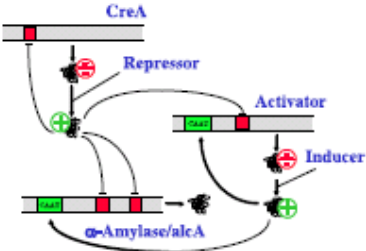
High level models



Low level models



- Global analysis (omics)
- ↓
- Data analysis (bioinformatics)
- ↓
- Reduction of dimensionality
- ↓
- System description**
- ↑
- Reaction/process kinetics
- ↑
- Molecular interactions

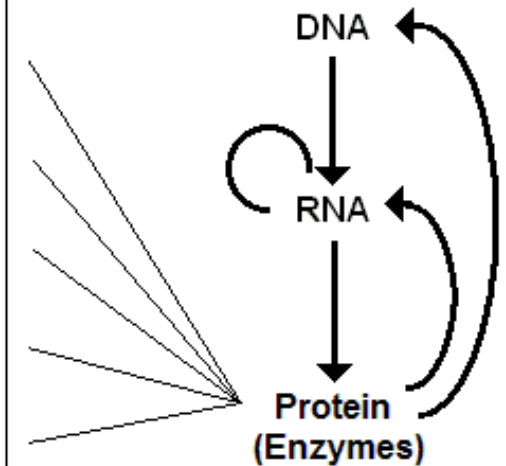


# Metabolic Engineering

-1	1	0	0	-1	1	0	0
0	-1	1	0	0	0	0	0
0	0	-1	1	0	0	0	0
0	0	0	-1	1	-1	1	0
0	0	-1	0	1	-1	0	1

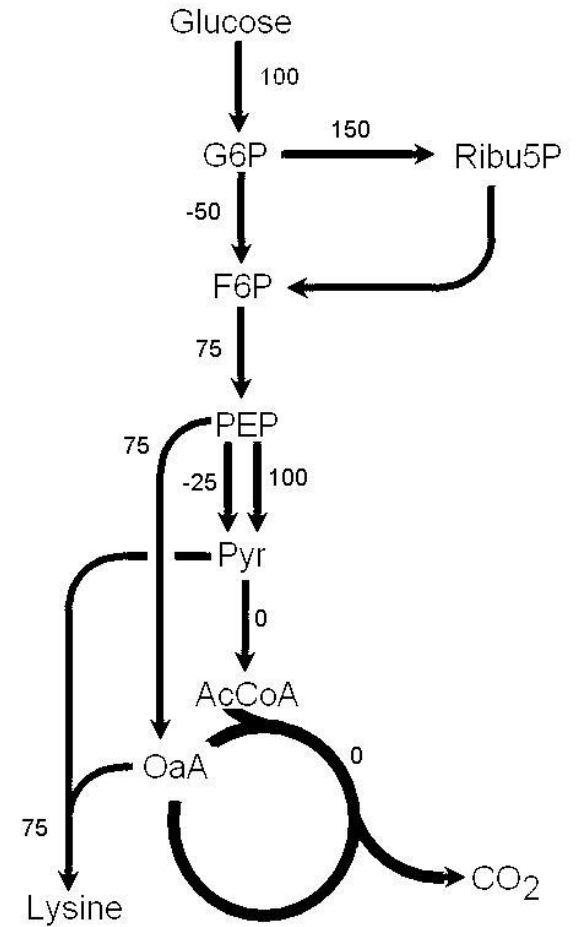
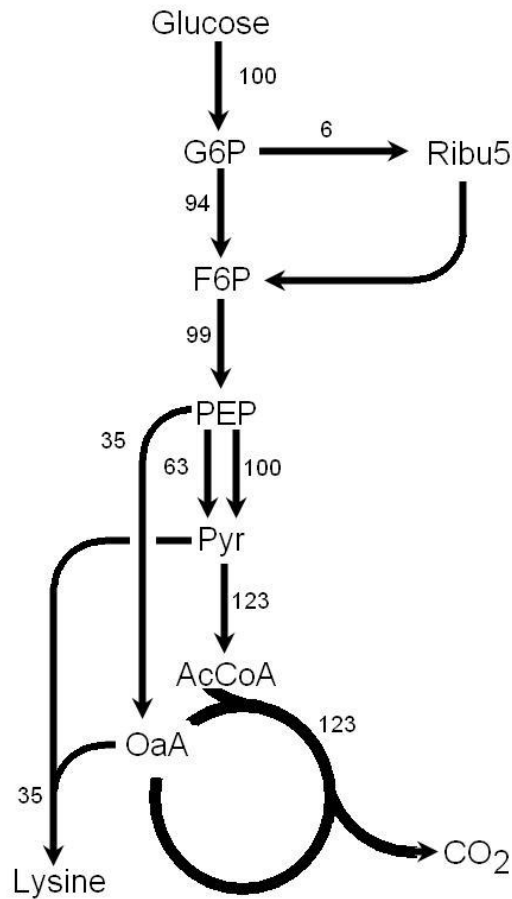
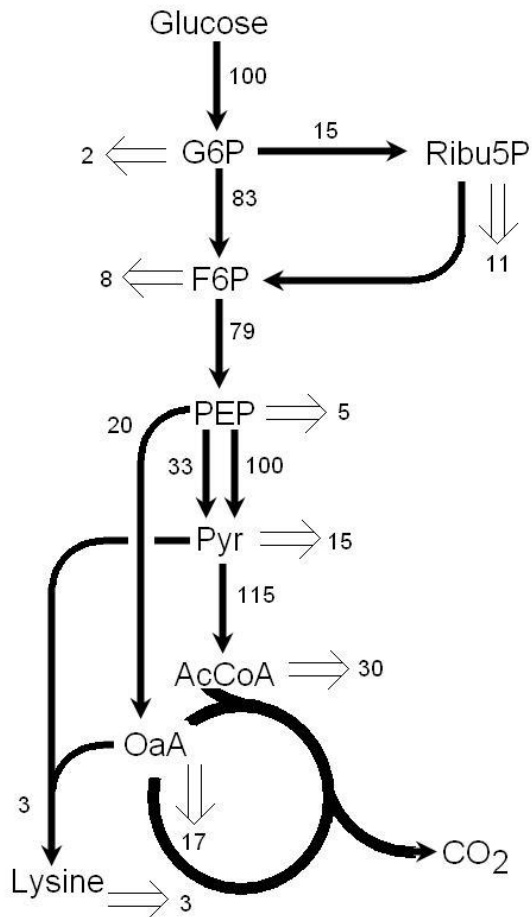
**Metabolic level  
regulation**

V1  
V2  
V3  
V4  
V5



**Hierarquical level  
regulation**

# Fluxos metabólicos

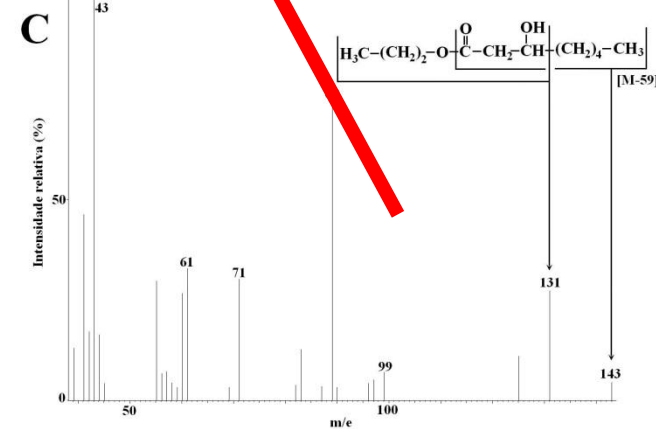
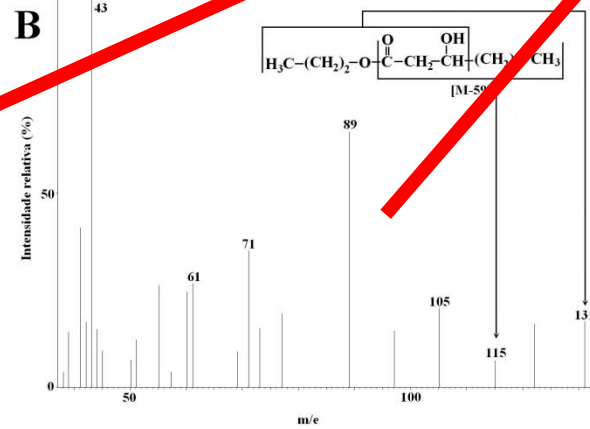
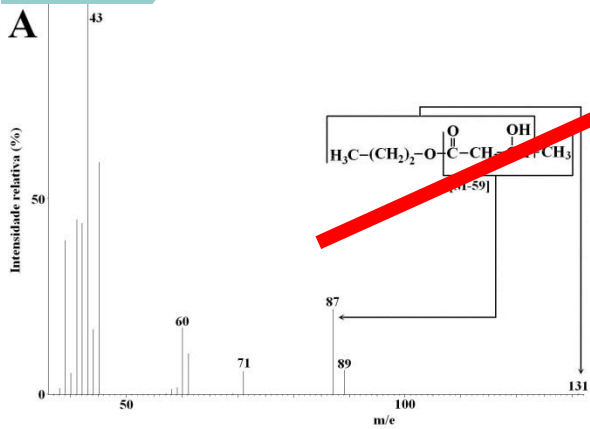


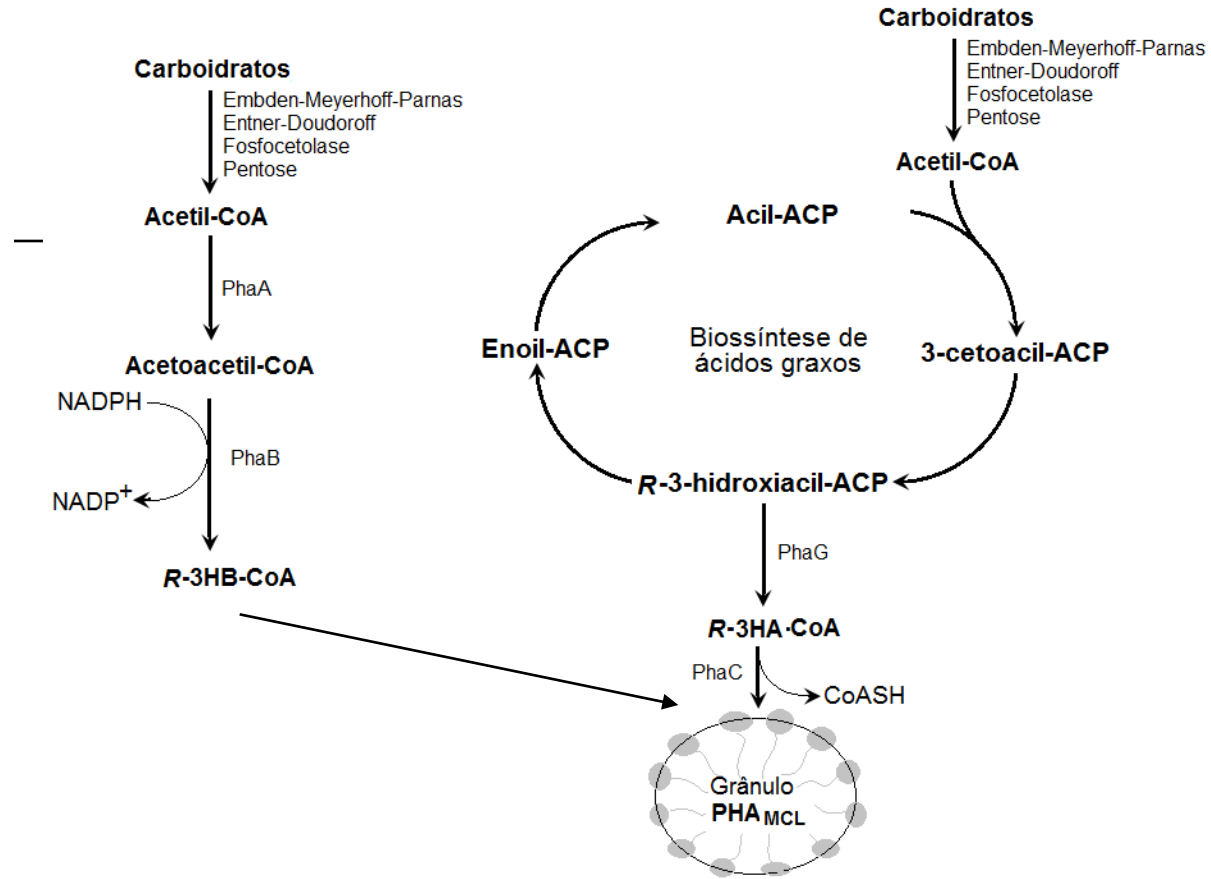
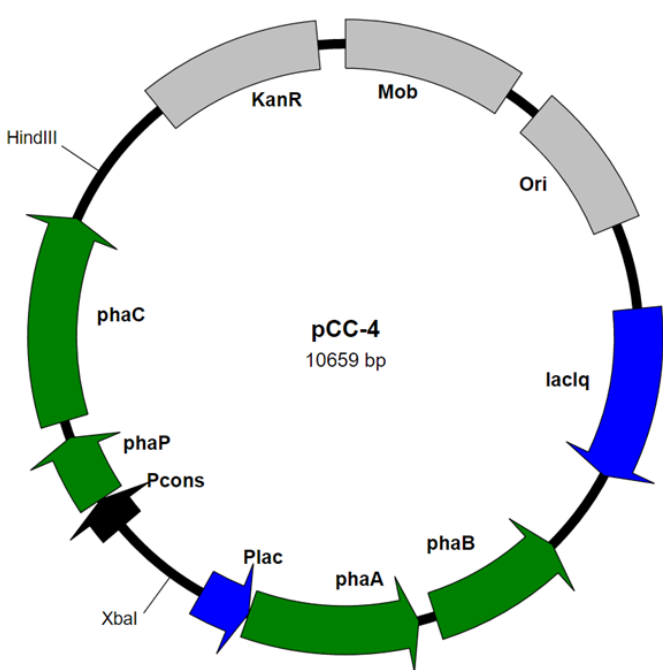
# *Pseudomonas* sp. abrigando genes de *R. eutropha*

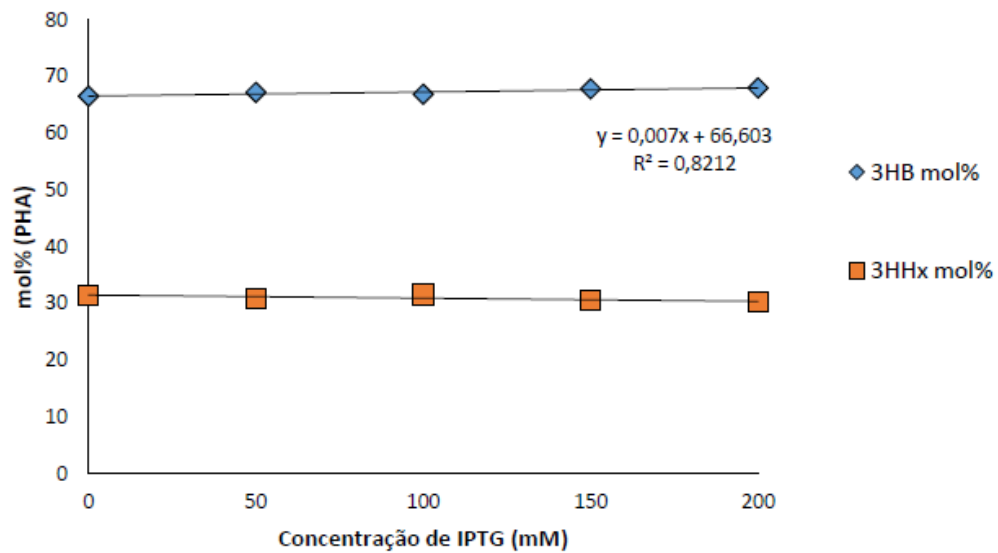
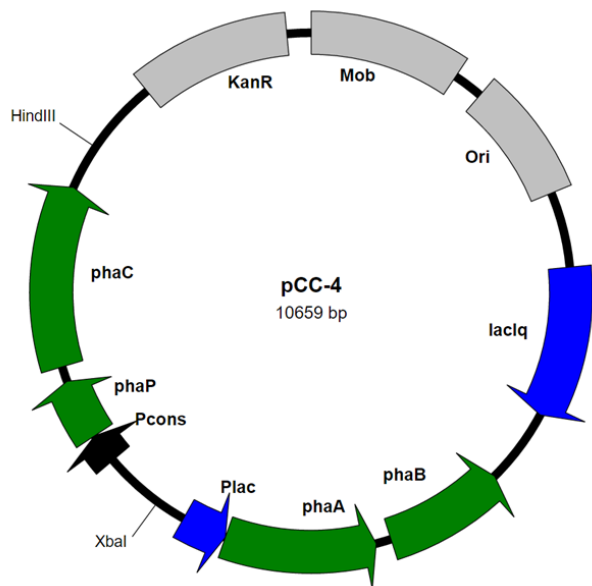
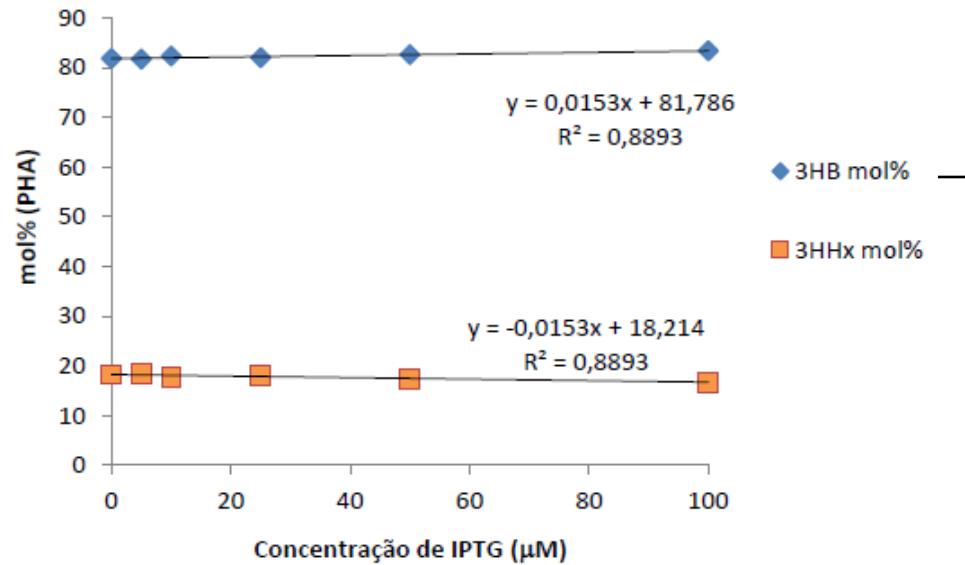
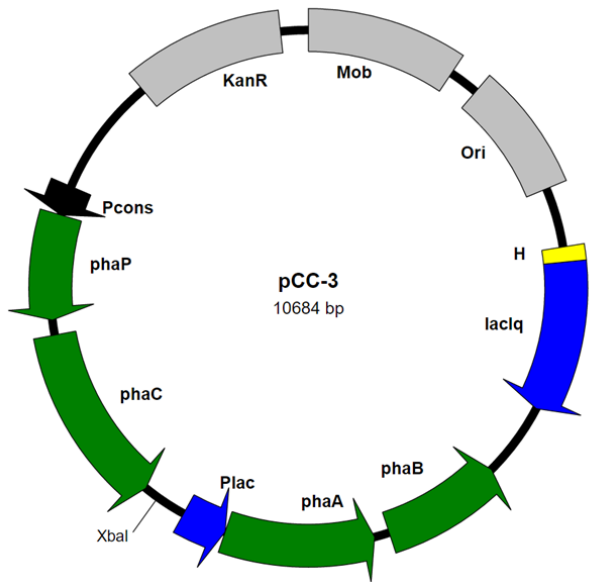
**Tabela 3.** Composição de PHA produzido. Análise de células liofilizadas e polímero purificado.

Linhagens recombinantes	Fonte de carbono	Material	PHA (mol%)				
			3HB	3HHx	3HO	3HD	3HDd
<i>Pseudomonas</i> sp. LFM046 pBBR1MCS-2	Octanoato	Cel. Liof.	0,71	13,69	73,72	9,95	1,34
		Polímero	0,86	14,35	75,14	9,65	Tr
<i>Pseudomonas</i> sp. LFM046 pBBR1MCS-2:: <i>phaB</i>	Octanoato	Cel. Liof.	0,00	15,13	57,02	12,10	15,76
		Polímero	2,75	20,52	74,99	1,73	Tr
<i>Pseudomonas</i> sp. LFM461 pBBR1MCS-2	Glicose	Cel. Liof.	0,00	0,00	0,00	47,99	52,01
		Polímero	-	-	-	-	-
<i>Pseudomonas</i> sp. LFM461 pBBR1MCS-2:: <i>phaC</i>	Glicose	Cel. Liof.	91,59	4,10	3,28	1,03	Tr
		Polímero	92,38	4,33	3,29	0,00	0,00

3HB – 3-hidroxibutirato      3HD – 3-hidroxidodecanoato      Polímero – polímero purificado.  
 3HHx – 3-hidroxihexanoato      3HDd – 3-hidroxidodecanoato      Cel. Liof. – células liofilizadas  
 3HO – 3-hidroxiocetanoato

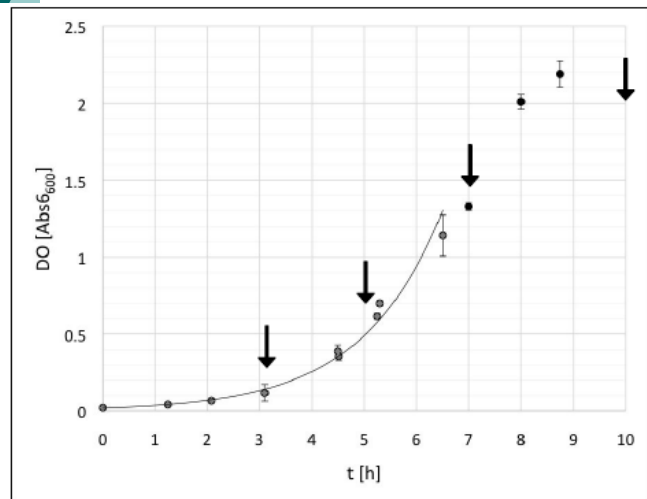
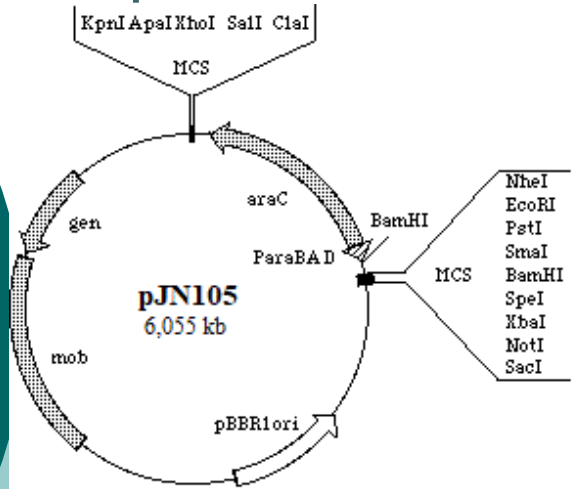




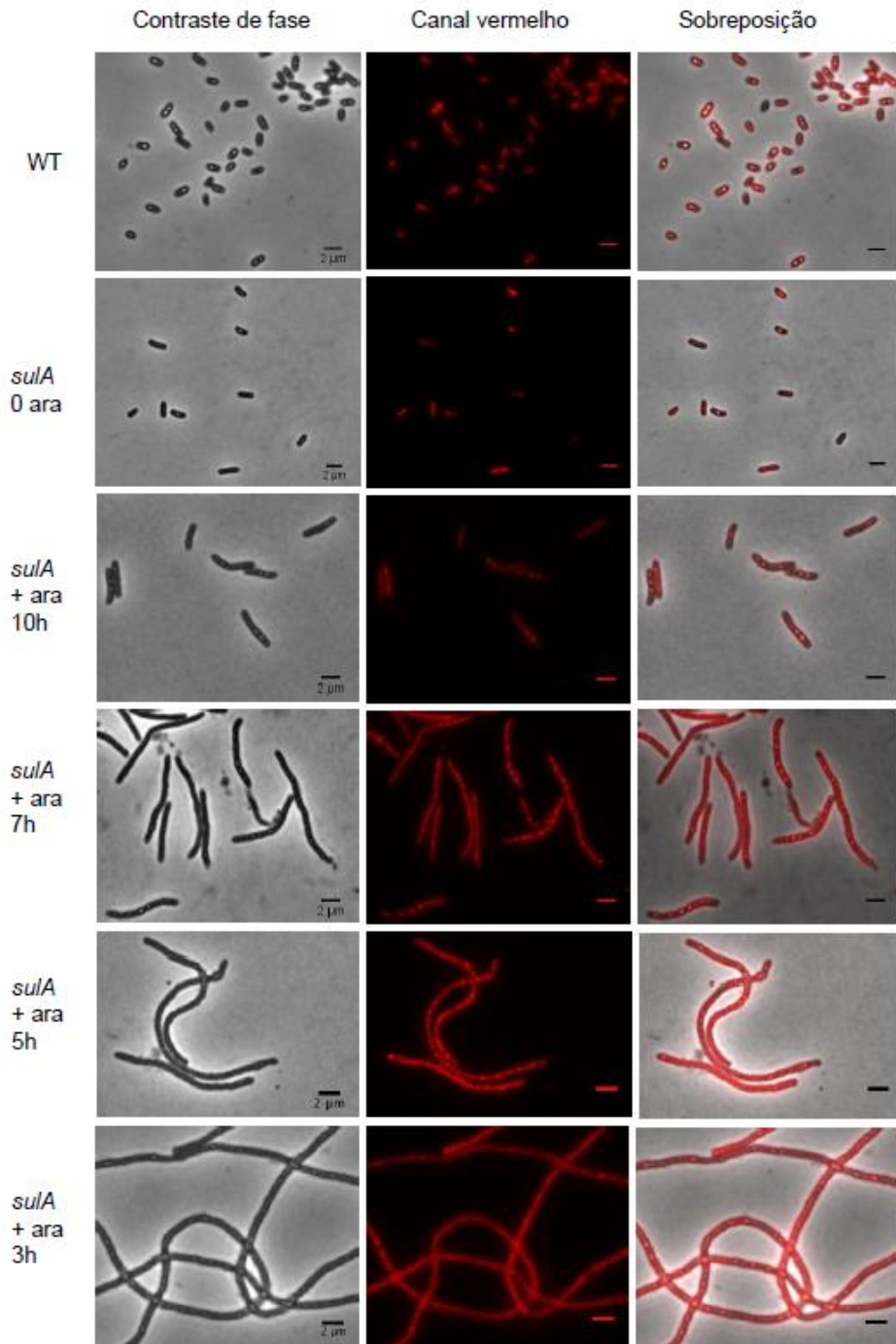


# *Pseudomonas* sp. LFM046

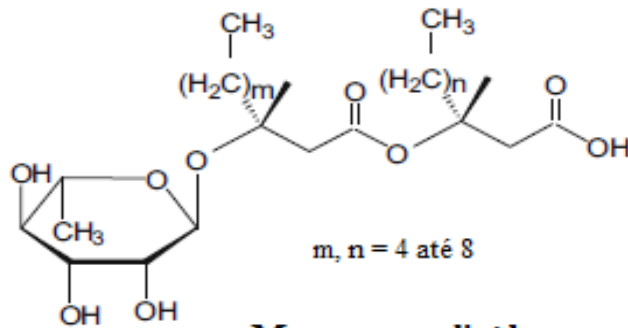
## pJN105::*sulA*



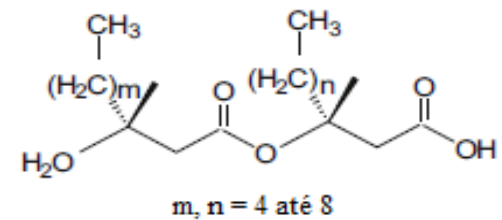
Imagens de microscopia de cultivos de *Pseudomonas* sp. LFM046:pJN105*sulA* com adição de L-arabinose em diferentes tempos de cultivo. As amostras foram retiradas após 72h de cultivo em meio para acúmulo de PHA com adição de L-arabinose (ara) 20 mM após 3h, 5h, 7h e 10h de cultivo ou sem adição, como indicado na figura. As amostras foram coradas com o corante de PHA Nile Red e analisadas em microscópio de fluorescência. Barra de escala de 2  $\mu$ m.



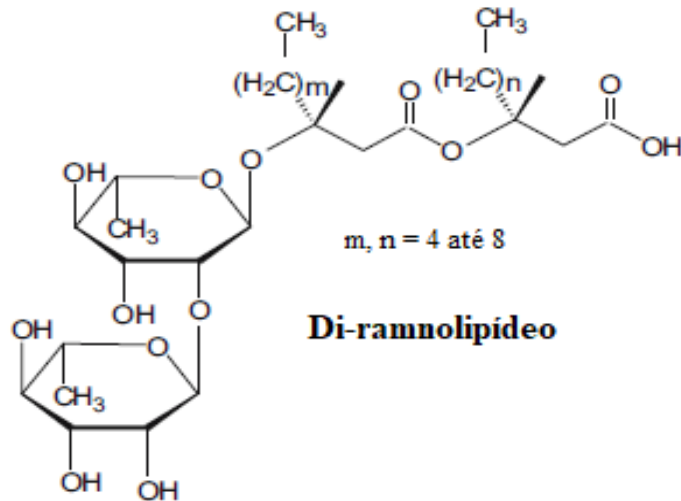
# Ramnolipídeos



**Mono-ramnolipídeo**



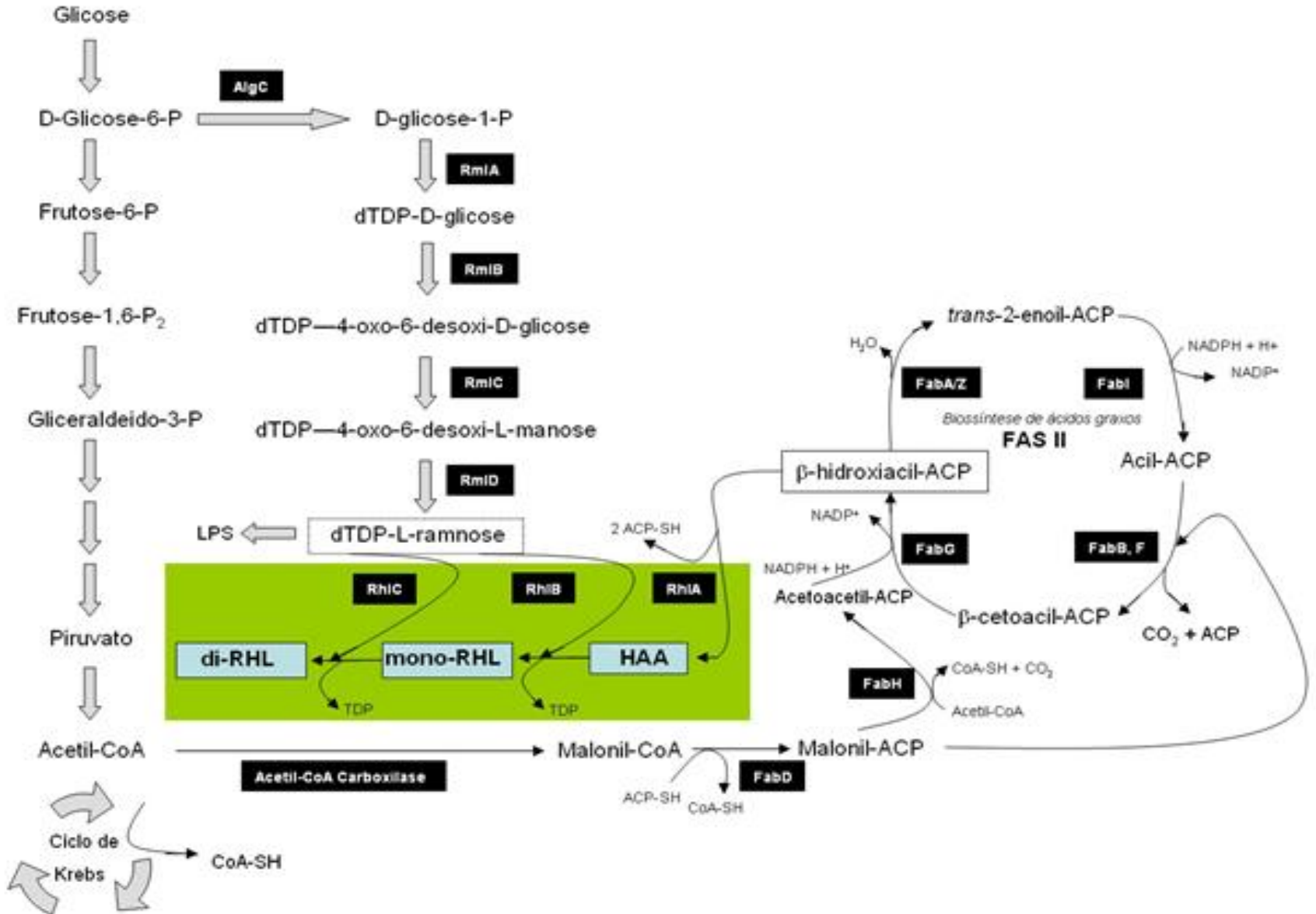
**3-(3-hidroxi)alcanoiloxi)alcanoato (HAA)**



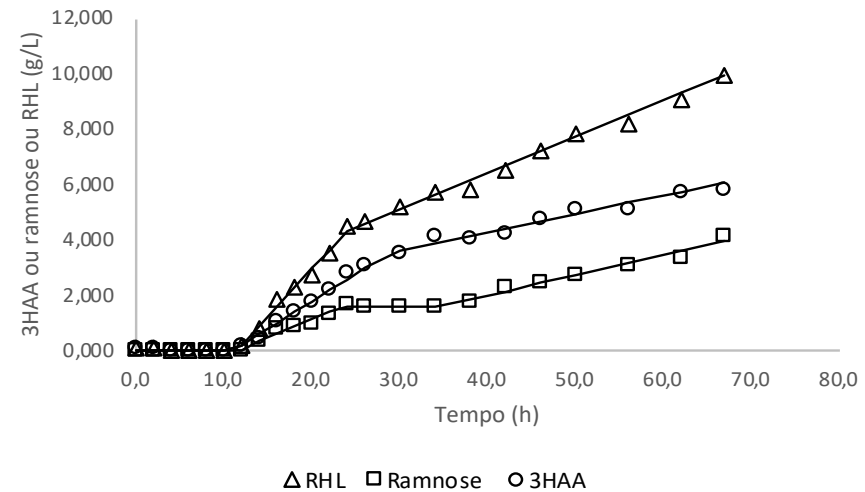
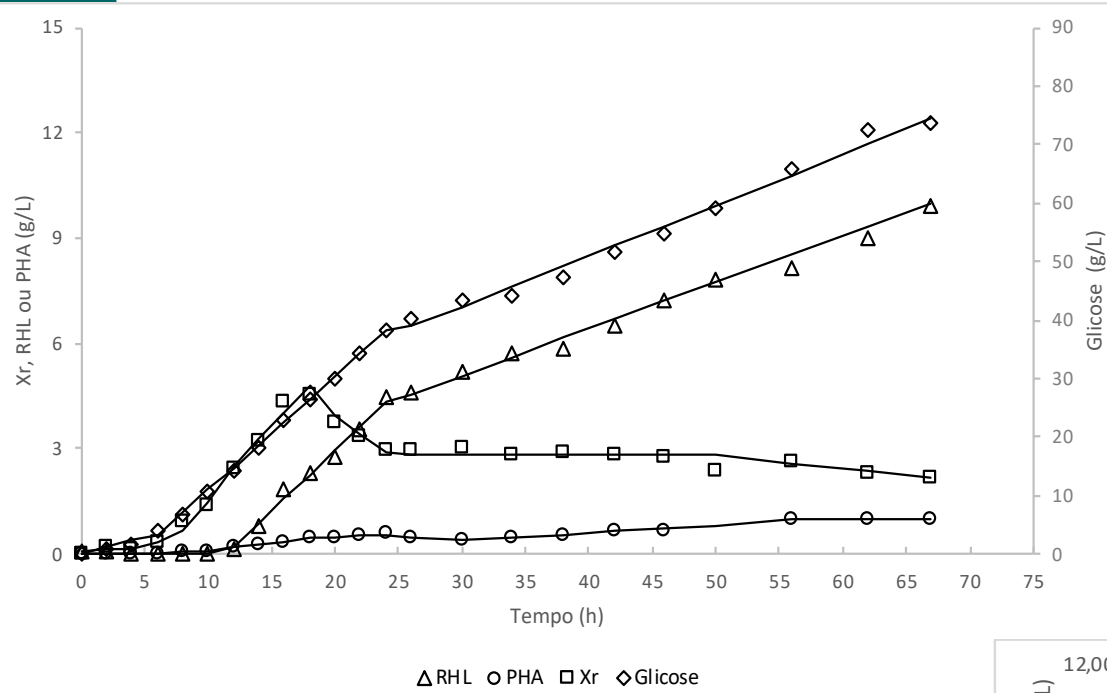
**Di-ramnolipídeo**



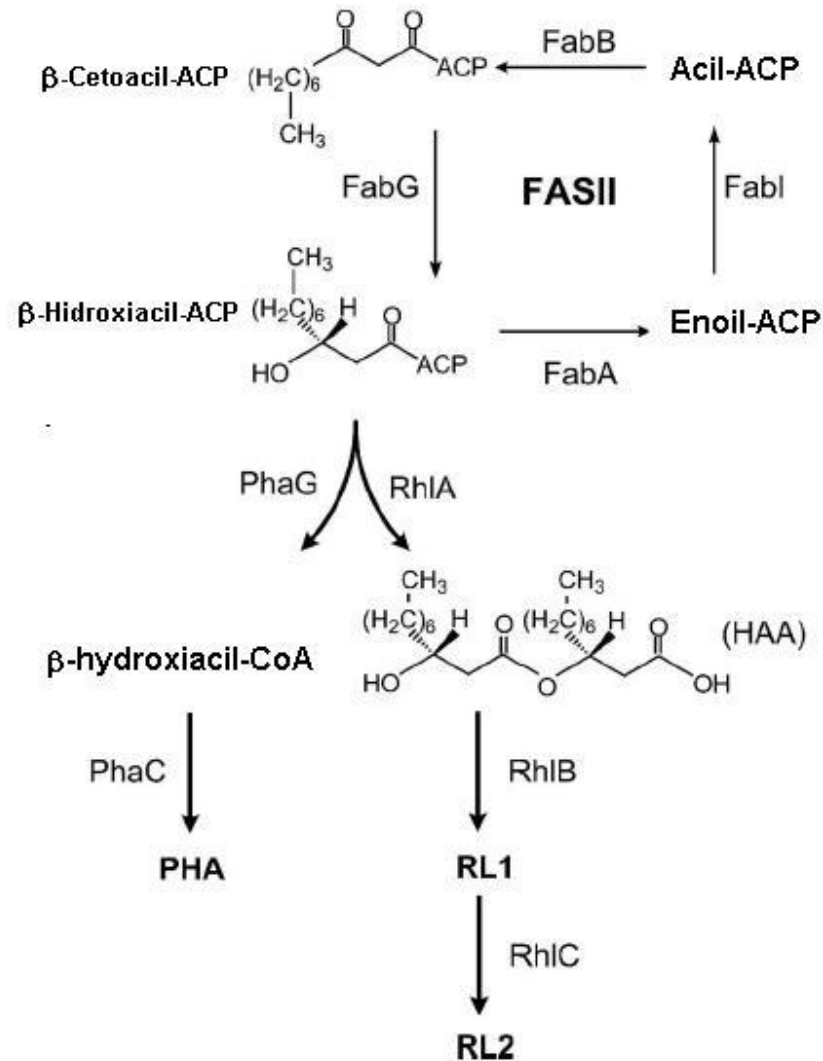
# Ramnolipídeos



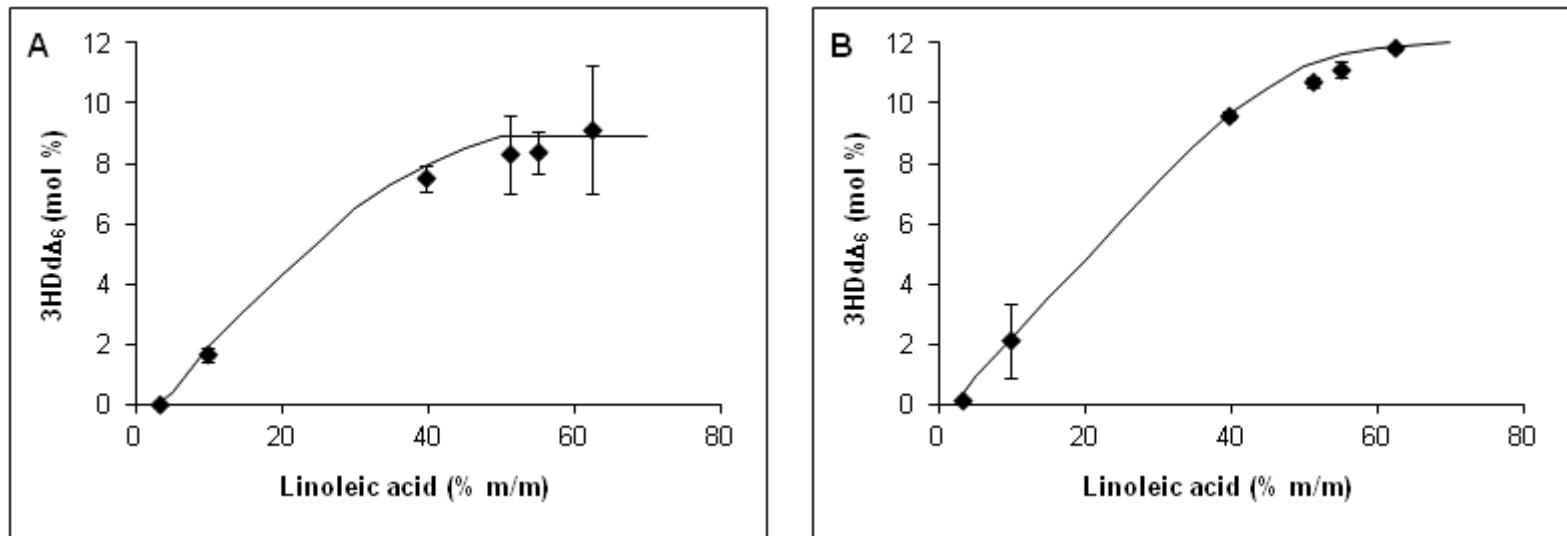
# Ramnolipídeos - Bioprocesso



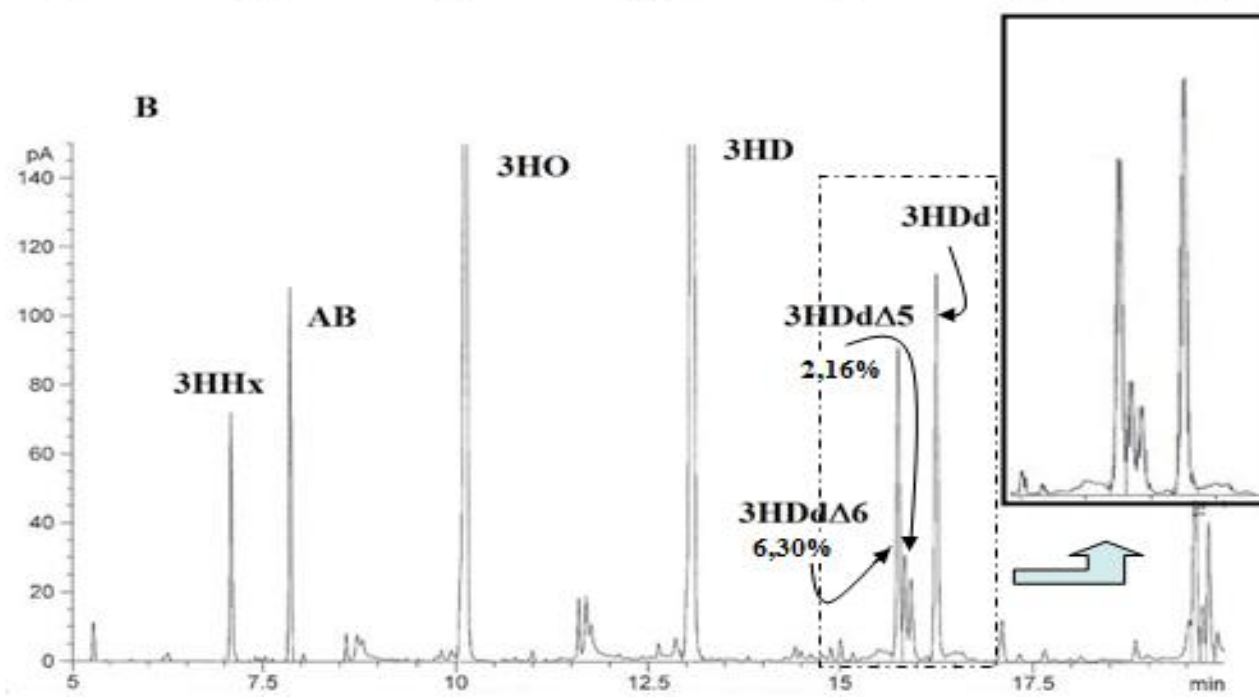
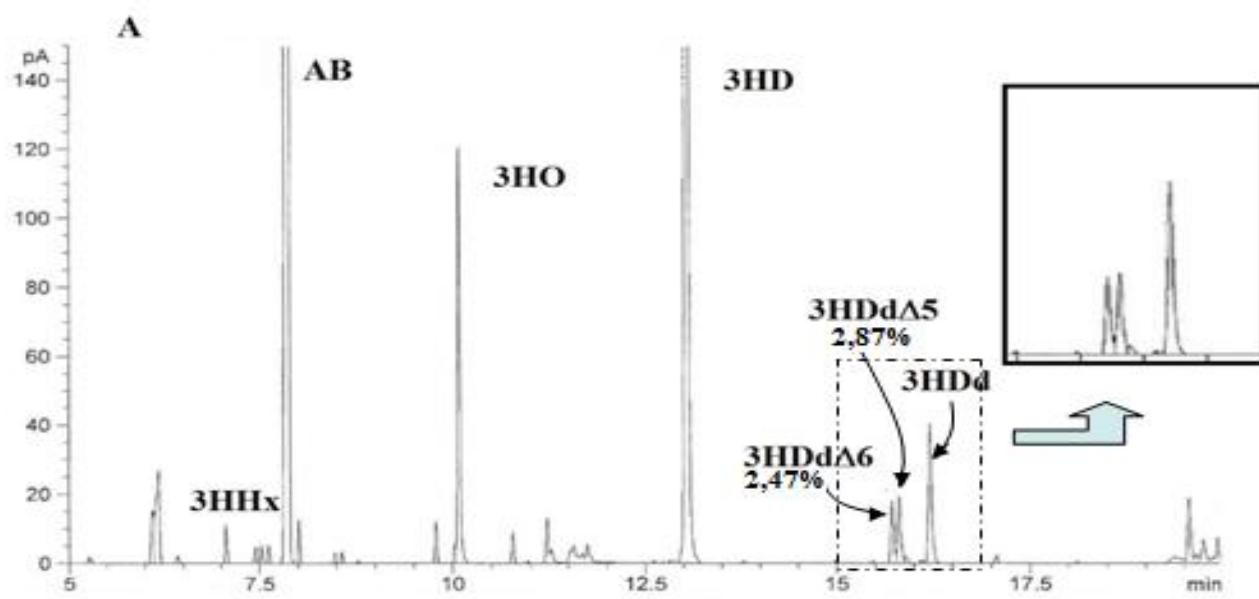
# Ramnolipídeos



# Ramnolipídios composição.



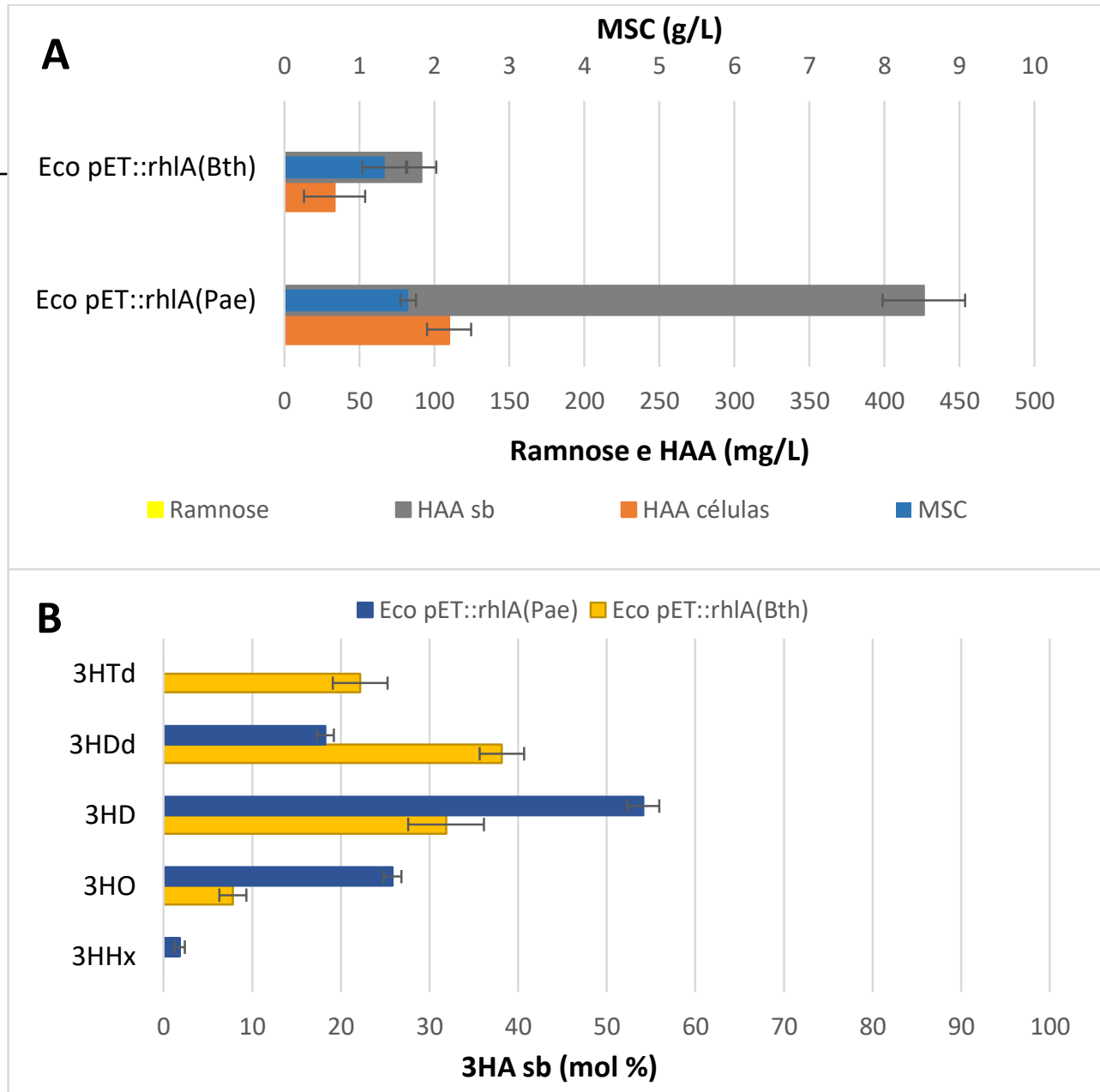
**Figure 2.** Relationship between linoleic acid fraction in plant oils supplied and 3-hydroxy-6-dodecenoic acid (3HDdD<sub>6</sub>) detected in rhamnolipids (A) or polyhydroxyalkanoates (B) produced by bacterial strain RMP1315.



# rhIA

linhagens	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q
<b>A</b> <i>P.aeruginosa</i> LFM634	100%																
<b>B</b> <i>P.aeruginosa</i> ATCC14886	100%	100%															
<b>C</b> <i>P.aeruginosa</i> LESB58	100%	100%	100%														
<b>D</b> <i>P.aeruginosa</i> M18	100%	100%	100%	100%													
<b>E</b> <i>P.aeruginosa</i> DK2	99%	99%	99%	99%	100%												
<b>F</b> <i>P.aeruginosa</i> PAO1	99%	99%	99%	99%	99%	100%											
<b>G</b> <i>P.aeruginosa</i> PA7	96%	96%	96%	96%	95%	95%	100%										
<b>H</b> <i>P.fluorescens</i> SBW25	65%	65%	65%	65%	65%	65%	65%	100%									
<b>I</b> <i>P.poa</i> RE1_1_14	62%	62%	62%	62%	62%	62%	64%	90%	100%								
<b>J</b> <i>P.chlororaphis</i> AGH13750	63%	63%	63%	63%	63%	63%	63%	60%	59%	100%							
<b>K</b> <i>B.mallei</i> ATCC23344	48%	48%	48%	48%	48%	48%	49%	43%	45%	45%	100%						
<b>L</b> <i>B.pseudomallei</i> 1710b	48%	48%	48%	48%	48%	48%	48%	43%	45%	45%	99%	100%					
<b>M</b> <i>B.thailandensis</i> E264	48%	48%	48%	48%	48%	48%	48%	42%	44%	45%	97%	97%	100%				
<b>N</b> <i>B.oklahomensis</i> C6786	46%	46%	46%	46%	46%	46%	47%	42%	42%	44%	93%	93%	92%	100%			
<b>O</b> <i>B.glumae</i> PG2	45%	45%	45%	45%	45%	45%	46%	42%	42%	43%	79%	79%	80%	78%	100%		
<b>P</b> <i>B.gladioli</i> BSR3	44%	44%	44%	44%	44%	44%	45%	42%	41%	43%	79%	79%	79%	80%	88%	100%	
<b>Q</b> <i>B.cenocepacia</i> J2315	44%	44%	44%	44%	44%	44%	45%	42%	42%	42%	76%	76%	77%	75%	75%	76%	100%

# *rhIA*



# Produção de 1,3-propanodiol

## Processo biotecnológico:

- Produção de propanodiol a partir do glicerol, por exemplo
- Bactérias produtoras como:  
*Klebsiella pneumoniae*, *Citrobacter freundii* e *Clostridium butyricum*



Vantagens:

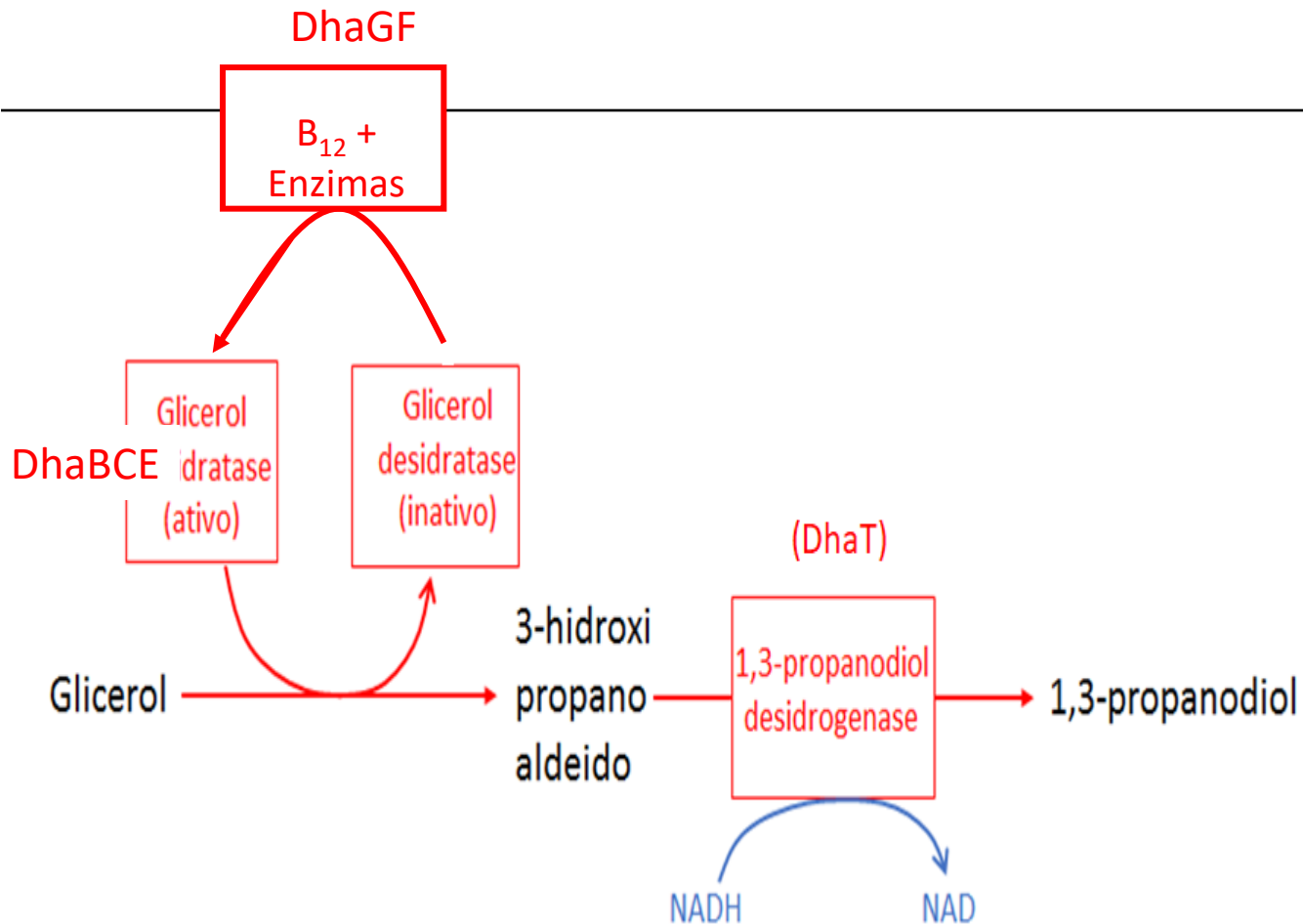
Fonte renovável, limpa e de baixo custo



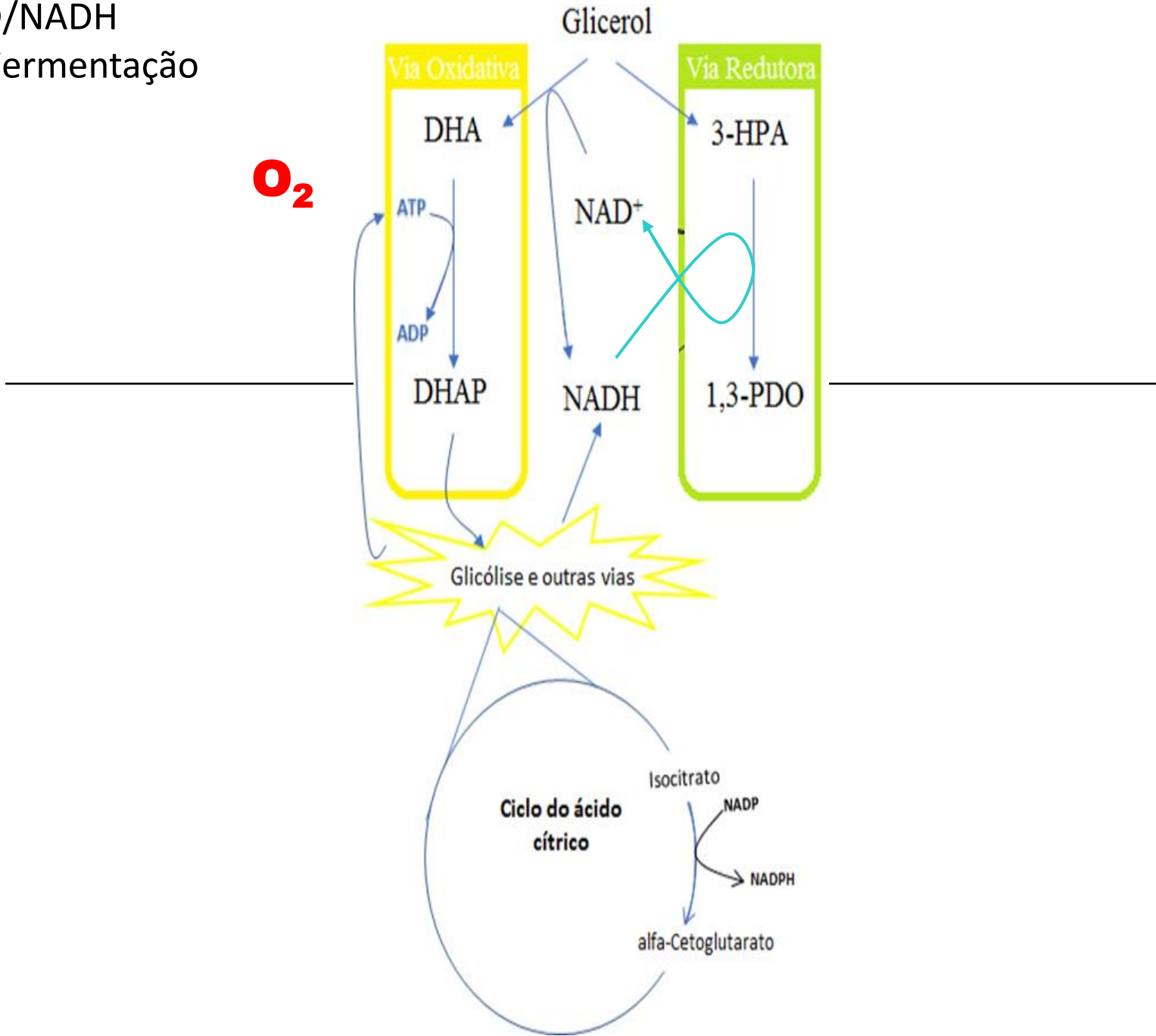


# Produção de 1,3-PDO por bactérias

- *Klebsiella pneumoniae*



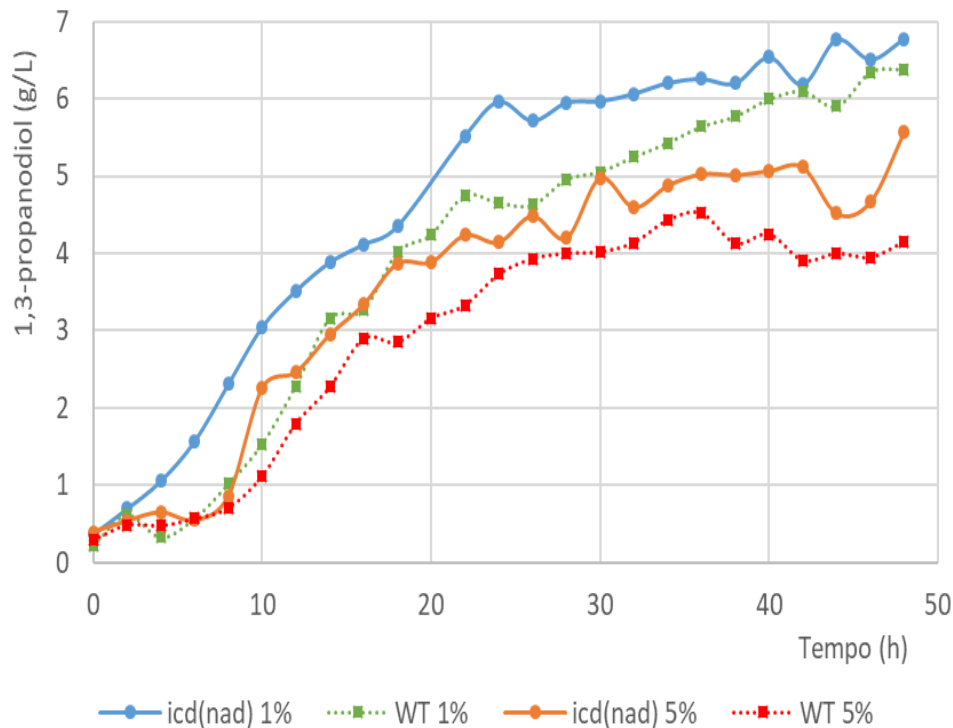
Balanco NAD/NADH  
Respiração/Fermentação



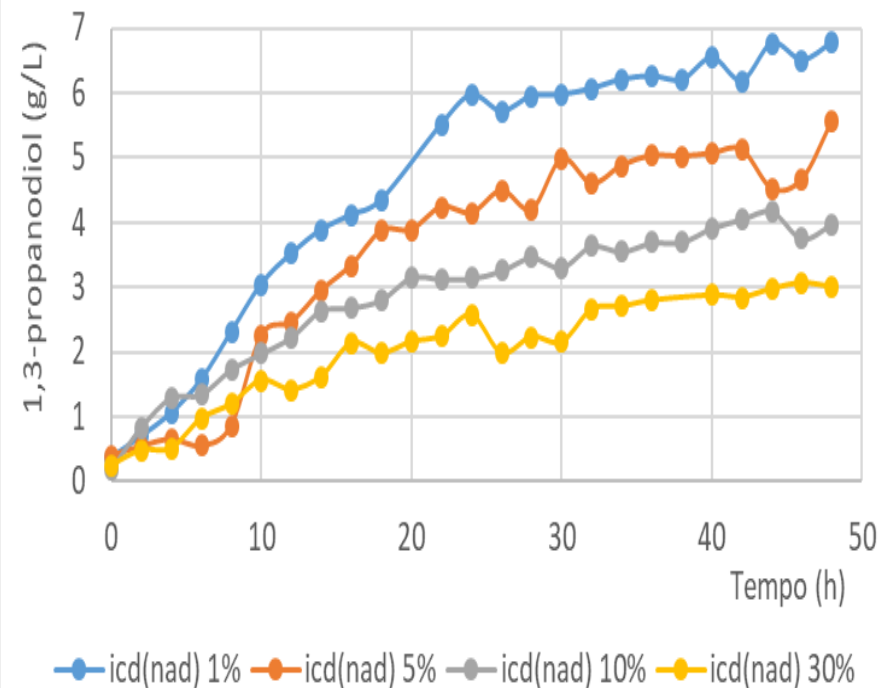
# Trabalhos anteriores

- *E. coli* transformada capaz de produzir 1,3-PDO

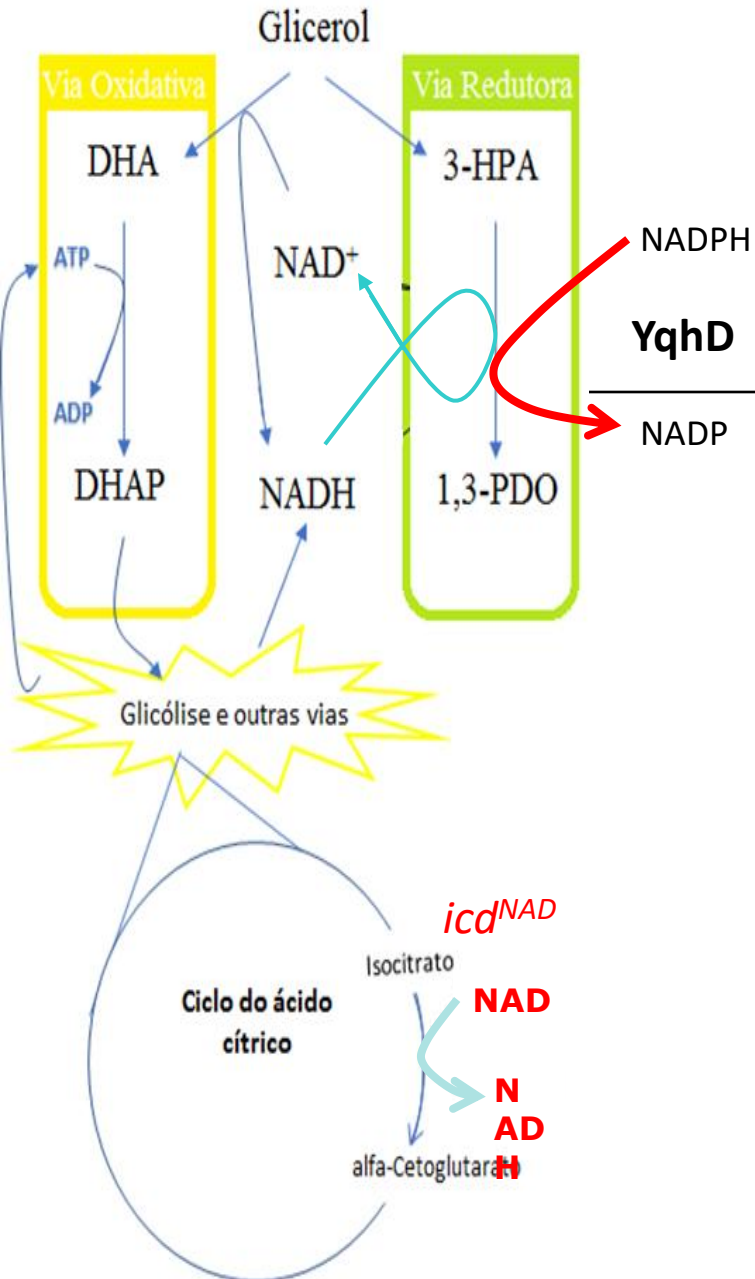
Produção de 1,3-PDO: *icd* WT e *icd* (*nad*)



Relação Oxigênio Dissolvido e produção de 1,3-PDO

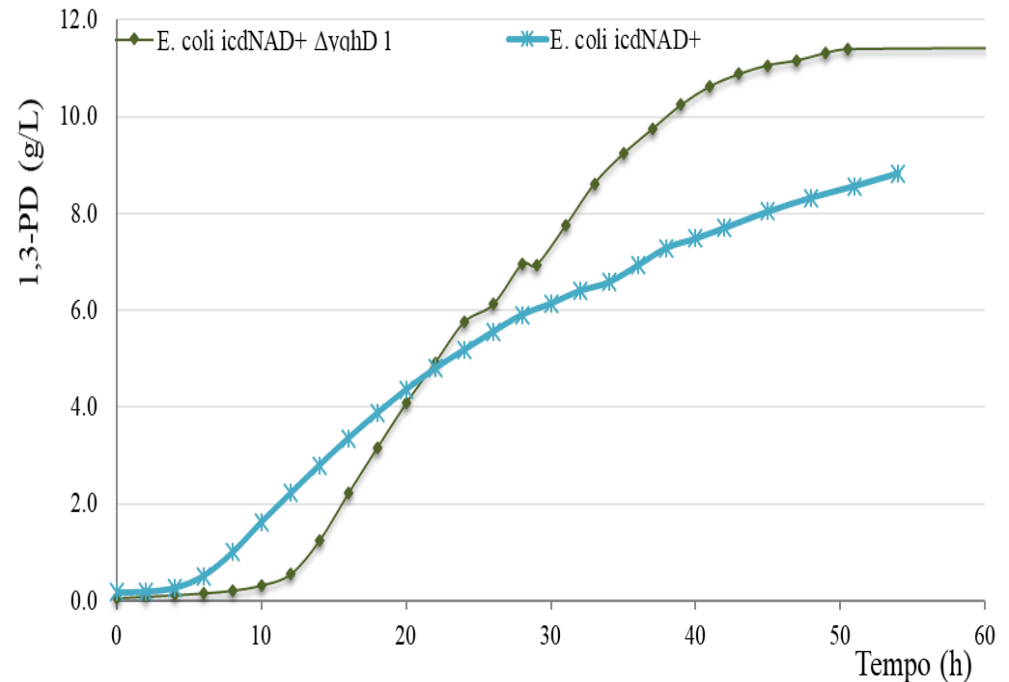


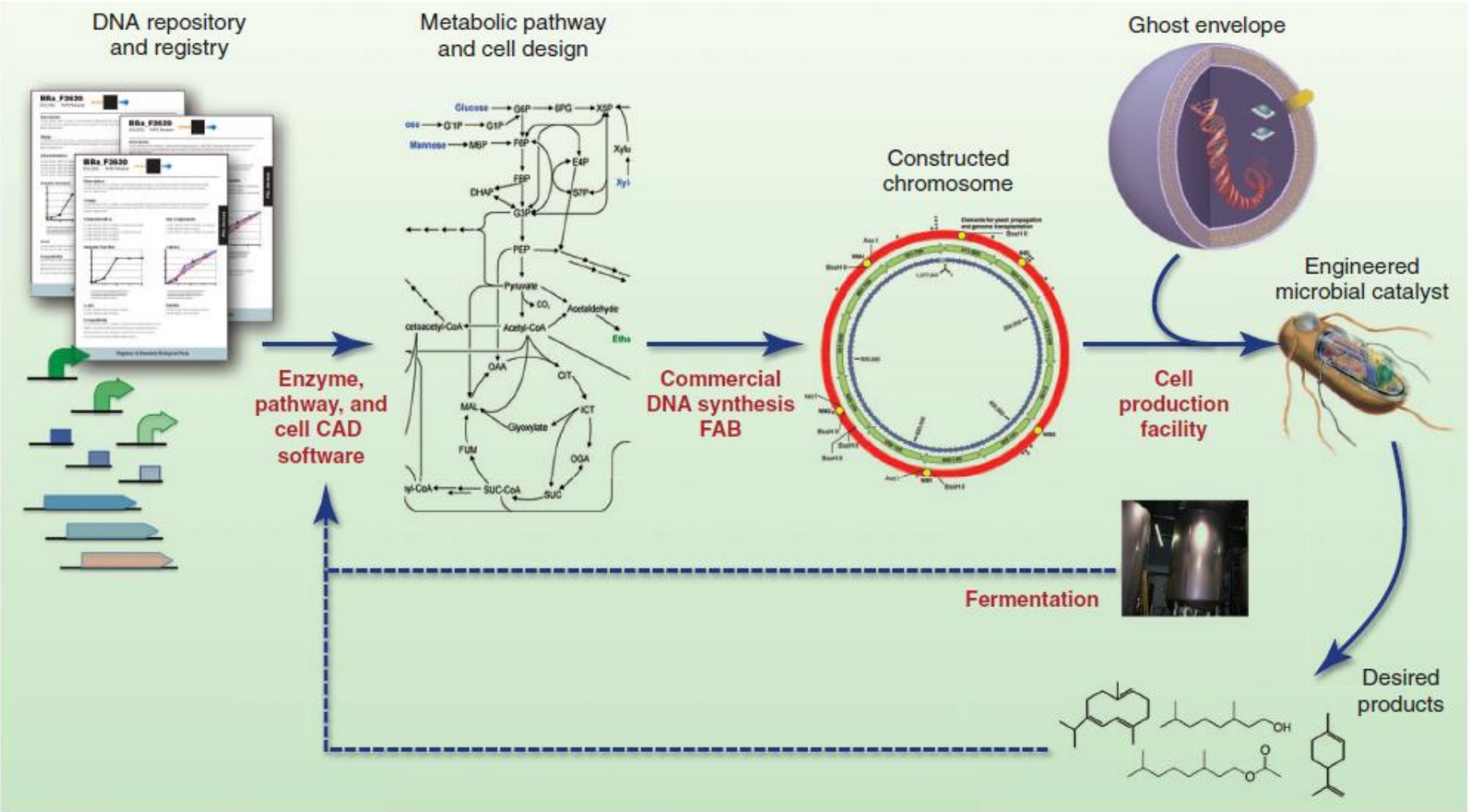
# Trabalhos anteriores



- O gene *yqhD* de *E. coli* é homólogo ao *dhaT*
- No entanto, YqhD é NADPH dependente.

*E. coli* MG1655 *icd<sup>NAD</sup> ΔyqhD* + pBBR1::*dha*





**Fig. 3.** The future of engineered biocatalysts. Pathways, enzymes, and genetic controls are designed from characteristics of parts (enzymes, promoters, etc.) by means of pathway and enzyme CAD software. The chromosomes encoding

those elements are synthesized at a FAB and incorporated into a ghost envelope to obtain the new catalyst. The design of the engineered catalyst is influenced by the desired product and the production process.

# Referências

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