

Modelos em bioprocessos

$$\mu_X = \frac{dX}{dt} \frac{1}{X}$$

$$Y_{X/S} = \frac{\mu_X}{\mu_S}$$

$$Y_{P/S} = \frac{\mu_P}{\mu_S}$$

Modelo de Monod

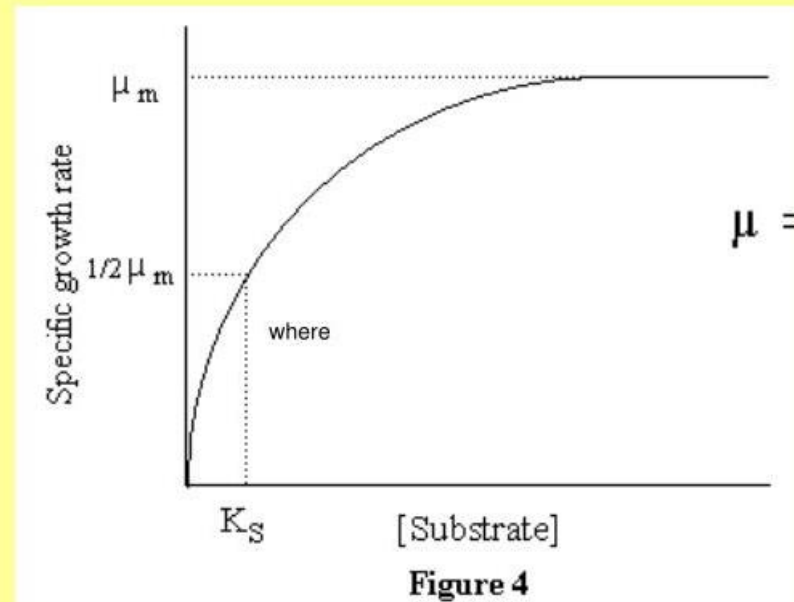


Figure 4

μ_m taxa específica de crescimento máxima
 K_s constante de saturação ou de Monod
 S concentração do substrato limitante .

Engenharia metabólica e Biologia de sistemas.

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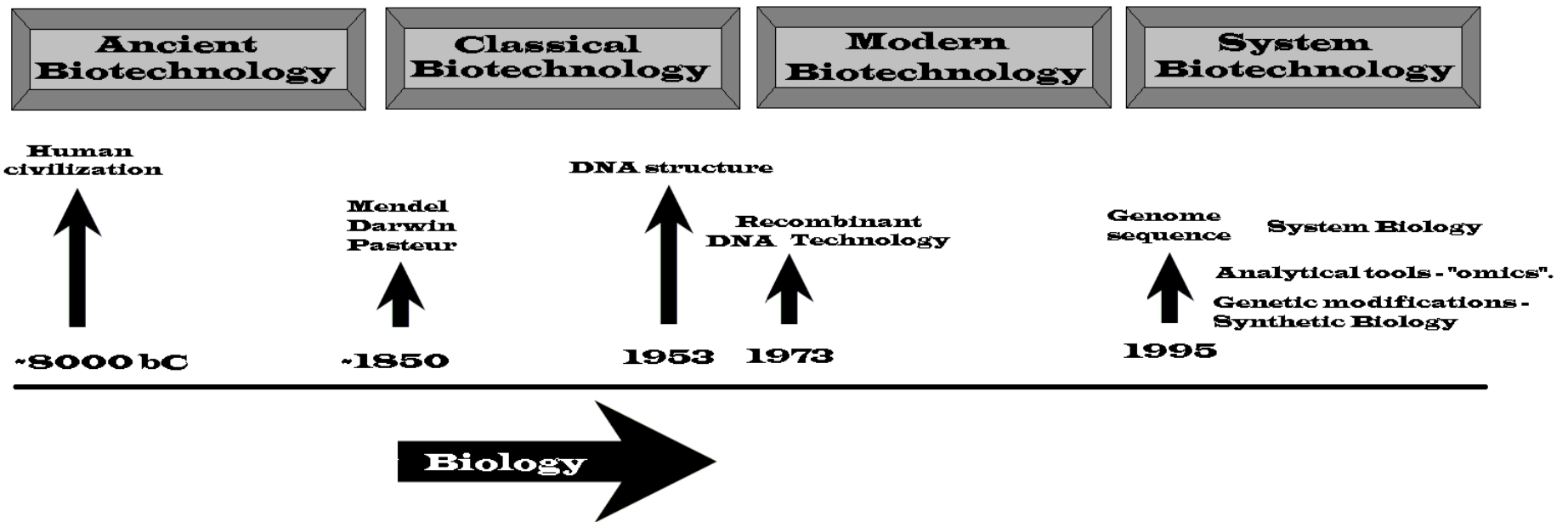


DEPARTAMENTO DE

MICroBiologia

UNIVERSIDADE DE SÃO PAULO

Biologia e Biotecnologia



ômicas

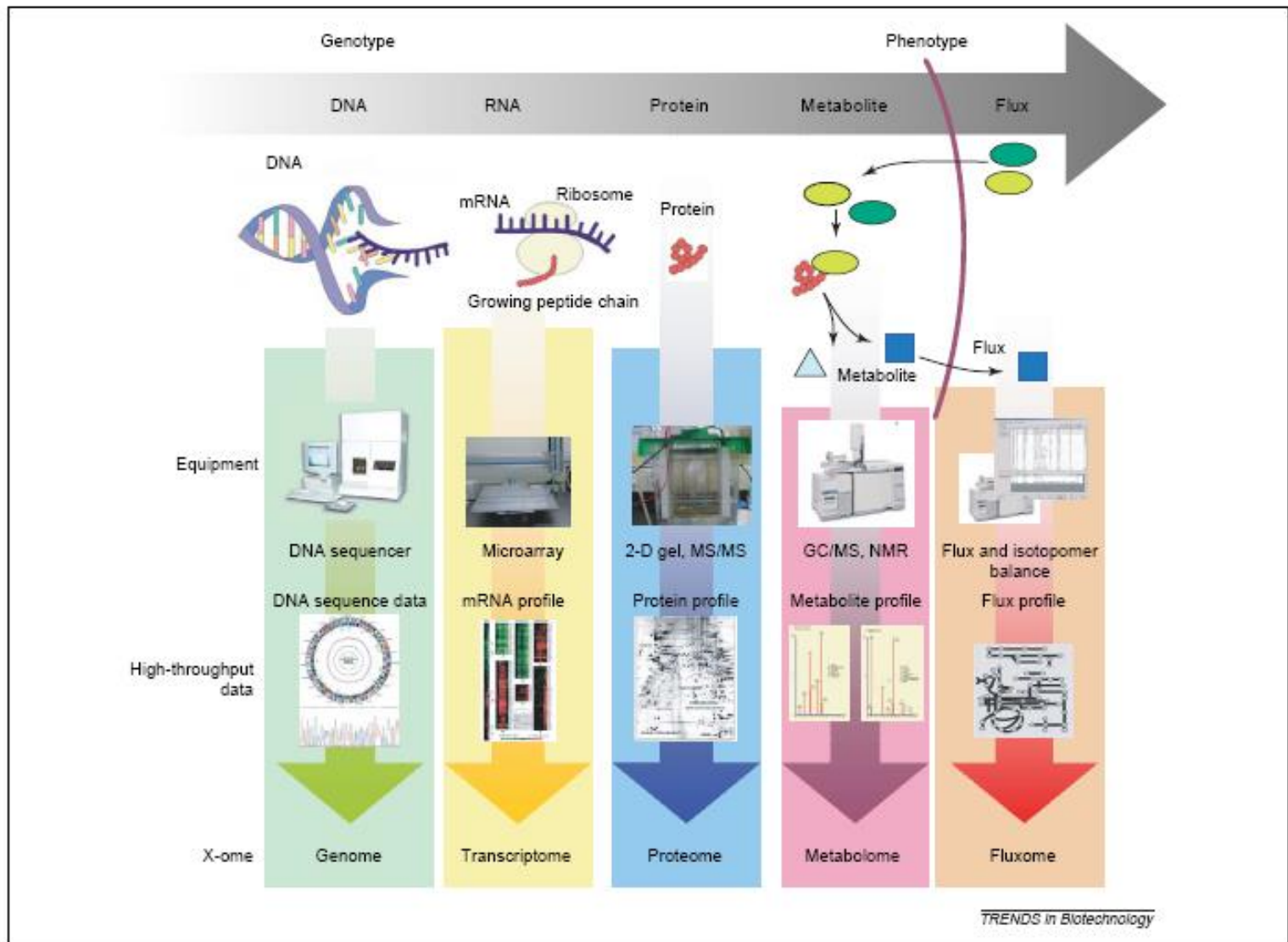
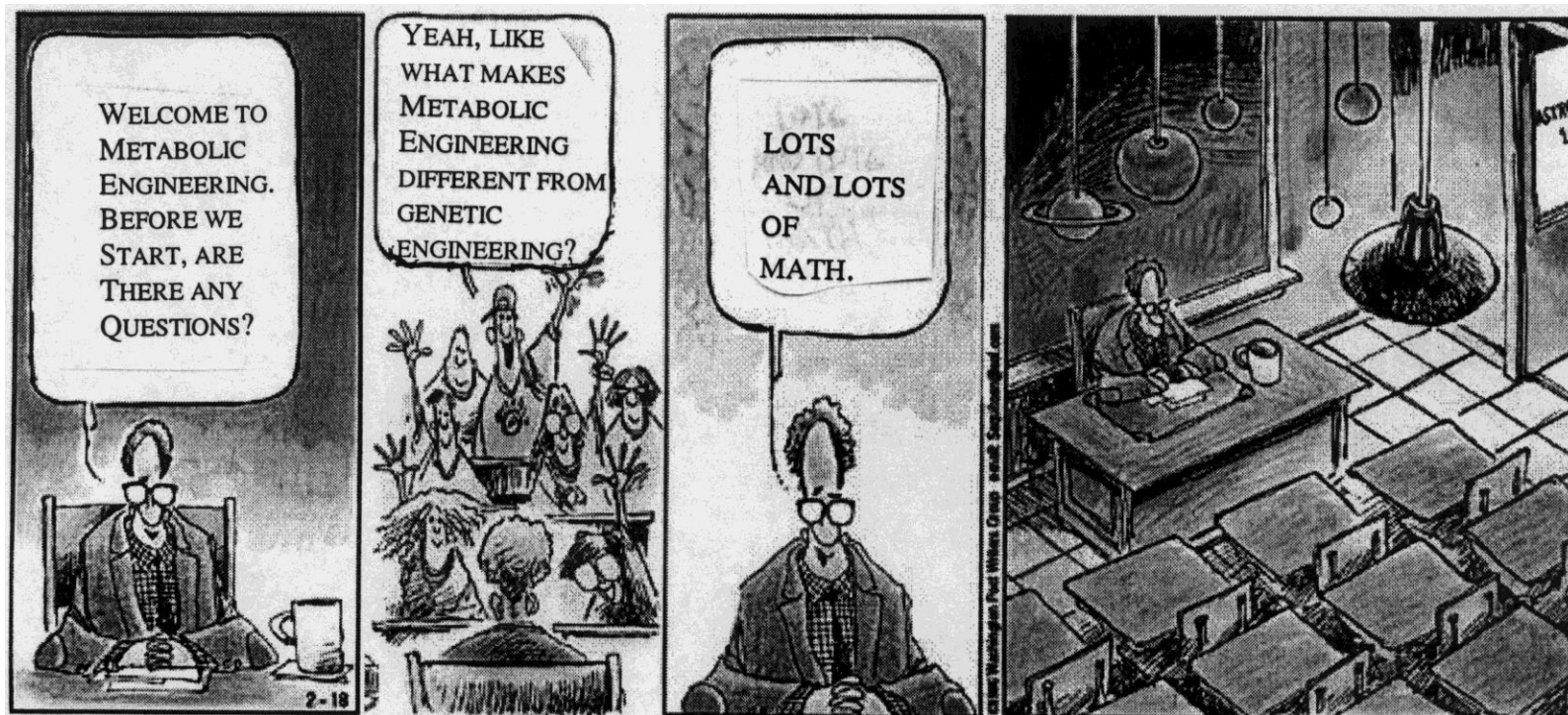


Figure 1. High-throughput omics research. Genomics advanced by the development of high-speed DNA sequencing is now accompanied by transcriptome profiling using DNA microarrays. Proteome profiling is joining the high-throughput race as 2D-gel electrophoresis combined with mass spectrometry is advancing. 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.

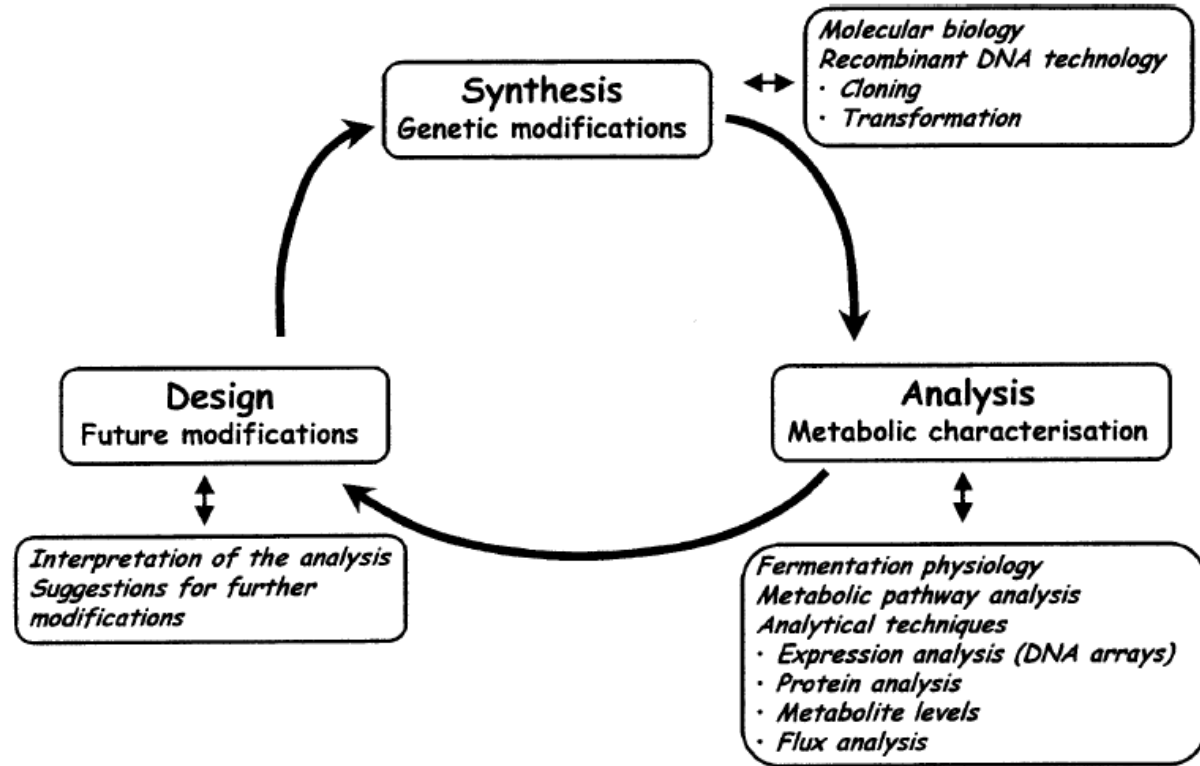


TABLE 1. Overview of reactions, metabolites, and ORFs in reconstructed metabolic networks^a

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 ^b

^a The reconstructed networks are described in references 6, 8, 17, and 18.

^b 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 ⁺	78
Glyceraldehyde-3P	13	NADPH	78
3-Phosphoglycerate	6	NADP ⁺	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).

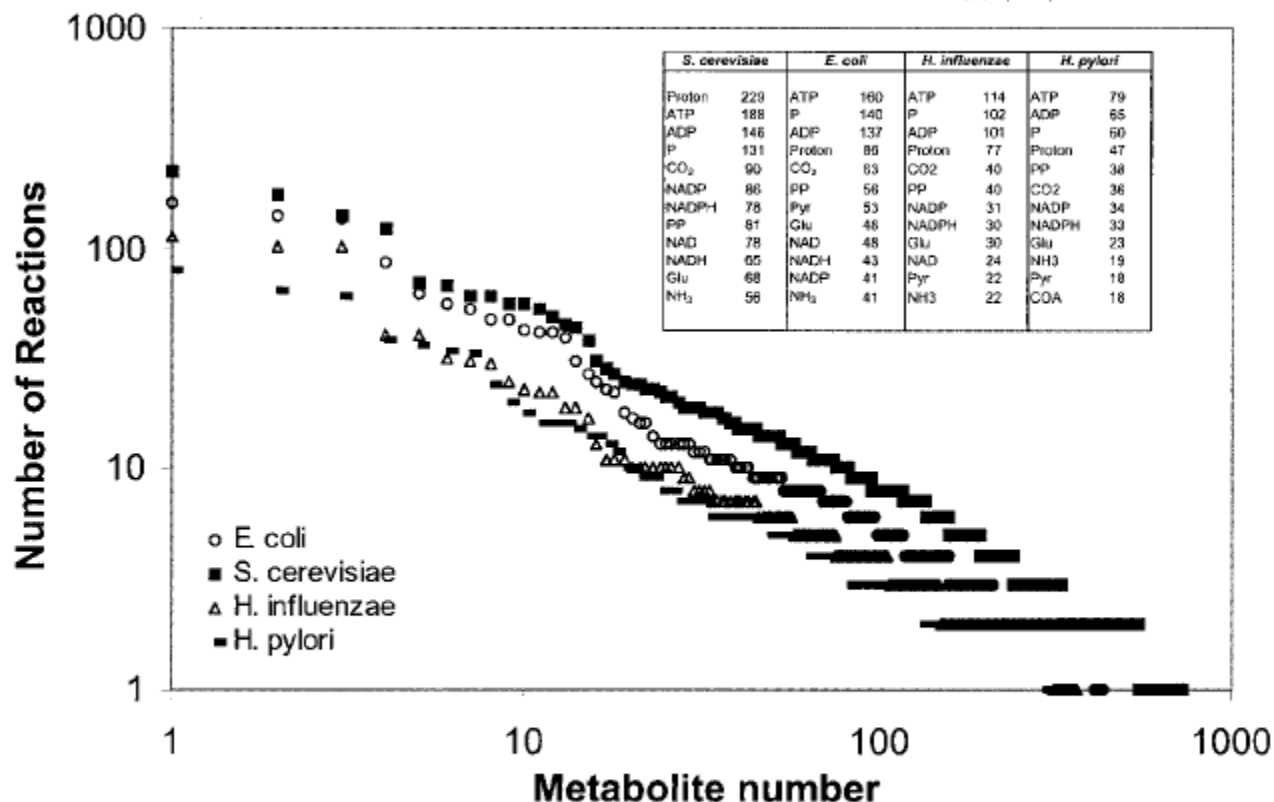
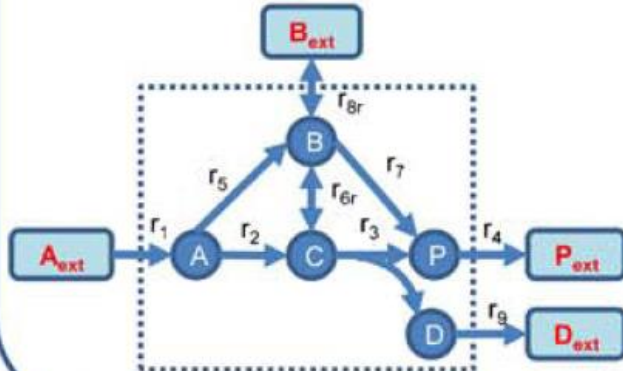


FIG. 1. Frequency plot of the number of reactions that each metabolite appears in for four different reconstructed metabolic networks. For each metabolic network the 10 metabolites that appear in the most reactions are listed. PP, pyrophosphate; COA, coenzyme A. The numbers in the box specify the numbers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

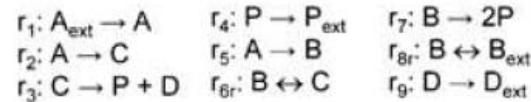
analysis of cellular metabolism

Problem statement

Network



Stoichiometric reactions



Stoichiometric matrix

$$\underline{S} = \begin{array}{c} \text{A} \\ \text{B} \\ \text{C} \\ \text{D} \\ \text{P} \end{array} \begin{array}{cccccccccc} r_1 & r_2 & r_3 & r_4 & r_5 & r_{6r} & r_7 & r_{8r} & r_9 \\ \begin{bmatrix} 1 & -1 & 0 & 0 & -1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & -1 & -1 & -1 & 0 \\ 0 & 1 & -1 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & -1 \\ 0 & 0 & 1 & -1 & 0 & 0 & 2 & 0 & 0 \end{bmatrix} \end{array}$$

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

Equations to solve

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

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

A

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

$\mu \cdot C$ (negligible)

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

$dC/dt = 0$ (steady state)

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

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).

Metabolic Flux Analysis

$$\underline{S}_u = \begin{matrix} & r_3 & r_4 & r_5 & r_{6r} & r_7 \\ \text{A} & 0 & 0 & -1 & 0 & 0 \\ \text{B} & 0 & 0 & 1 & -1 & -1 \\ \text{C} & -1 & 0 & 0 & 1 & 0 \\ \text{D} & 1 & 0 & 0 & 0 & 0 \\ \text{P} & 1 & -1 & 0 & 0 & 2 \end{matrix}$$

$$\underline{S}_m = \begin{matrix} & r_1 & r_2 & r_{8r} & r_9 \\ \text{A} & 1 & -1 & 0 & 0 \\ \text{B} & 0 & 0 & -1 & 0 \\ \text{C} & 0 & 1 & 0 & 0 \\ \text{D} & 0 & 0 & 0 & -1 \\ \text{P} & 0 & 0 & 0 & 0 \end{matrix}$$

Measured fluxes

$$\underline{r}_m = \begin{bmatrix} r_1 \\ r_2 \\ r_{8r} \\ r_9 \end{bmatrix} = \begin{bmatrix} 1 \\ 0.3 \\ 0 \\ 0.75 \end{bmatrix}$$

Equations to solve

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

$$\begin{bmatrix} \underline{S}_u & \underline{S}_m \end{bmatrix} \begin{bmatrix} \underline{r}_u \\ \underline{r}_m \end{bmatrix} = \underline{0}$$

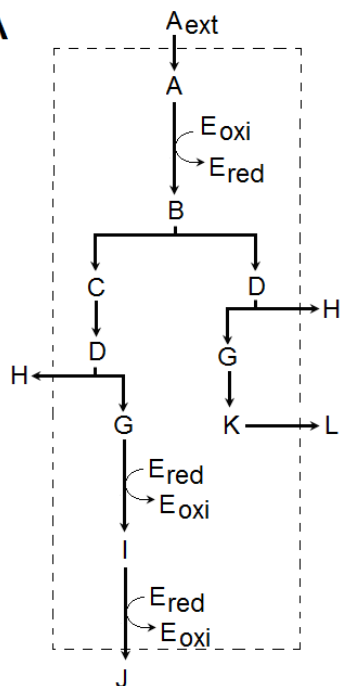
$$\underline{r}_u = -\underline{S}_u^{-1} \cdot \underline{S}_m \cdot \underline{r}_m$$

Solution

$$\underline{r}_u = \begin{bmatrix} r_3 \\ r_4 \\ r_5 \\ r_{6r} \\ r_7 \end{bmatrix} = \begin{bmatrix} 0.75 \\ 1.25 \\ 0.7 \\ 0.45 \\ 0.25 \end{bmatrix}$$

B

A



B

	r_1	r_2	r_3	r_4	r_5	r_6	r_7	r_8	r_9
Aext	-1	0	0	0	0	0	0	0	0
A	1	-1	0	0	0	0	0	0	0
B	0	1	-1	0	0	0	0	0	0
C	0	0	1	-1	0	0	0	0	0
D	0	0	1	1	-1	0	0	0	0
Eoxi	0	-1	0	0	0	1	1	0	0
Ered	0	1	0	0	0	-1	-1	0	0
G	0	0	0	0	1	-1	0	-1	0
H	0	0	0	0	1	0	0	0	0
I	0	0	0	0	0	1	-1	0	0
J	0	0	0	0	0	0	1	0	0
K	0	0	0	0	0	0	0	1	-1
L	0	0	0	0	0	0	0	0	1

C

$$\begin{bmatrix} -1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & -1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & -1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & -1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & -1 & 0 & 0 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & -1 & -1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & -1 & 0 & -1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & -1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \times \begin{bmatrix} r_1 \\ r_2 \\ r_3 \\ r_4 \\ r_5 \\ r_6 \\ r_7 \\ r_8 \\ r_9 \end{bmatrix} = \begin{bmatrix} dA/dt \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ dH/dt \\ 0 \\ dJ/dt \\ 0 \\ dL/dt \end{bmatrix}$$

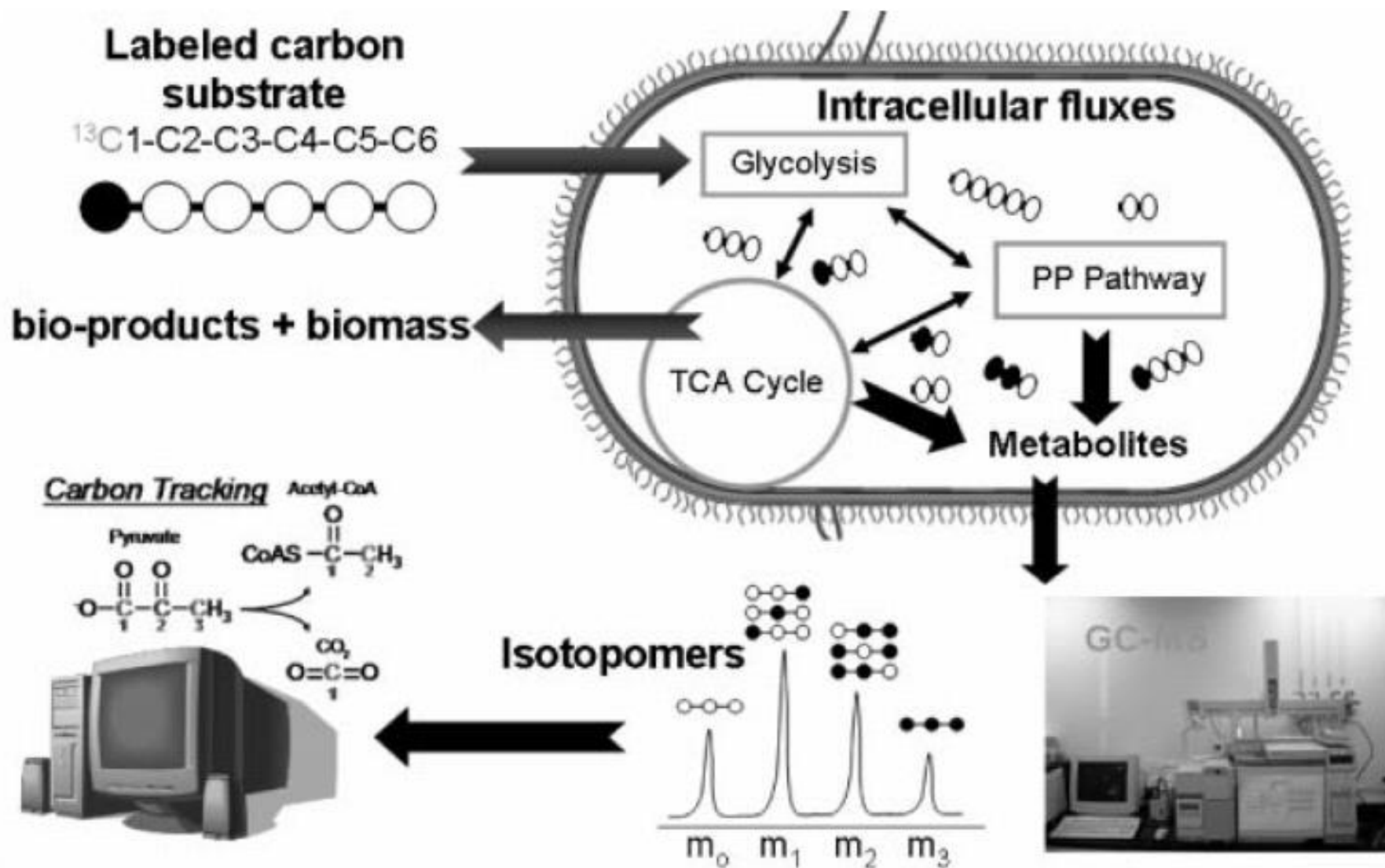


FIGURE 2. Protocol for ^{13}C -based flux analysis. [Color figure can be viewed in the online issue, which is available at www.interscience.wiley.com.]

Metabolic Engineering

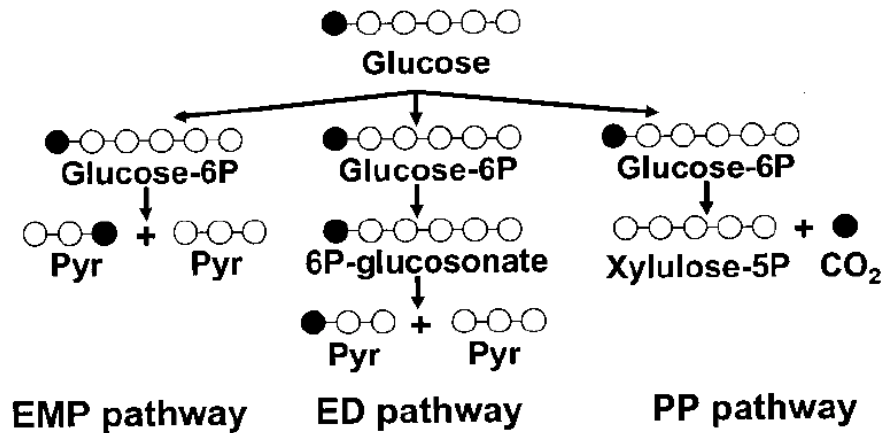


FIG. 2. Illustration of how measurement of the ^{13}C enrichment patterns can be used to identify active pathways. EMP, Embden-Mey-erhof-Parnas; ED, Entner-Doudoroff; PP, pentose phosphate.

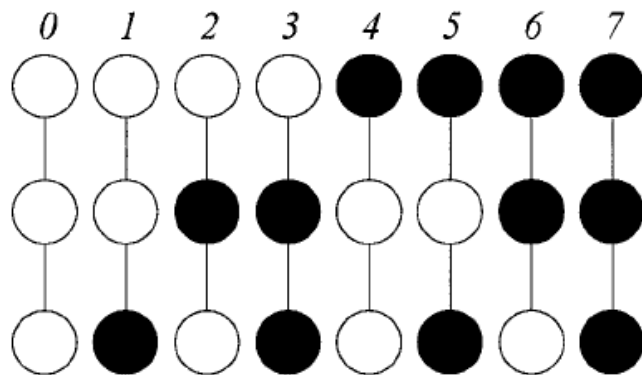
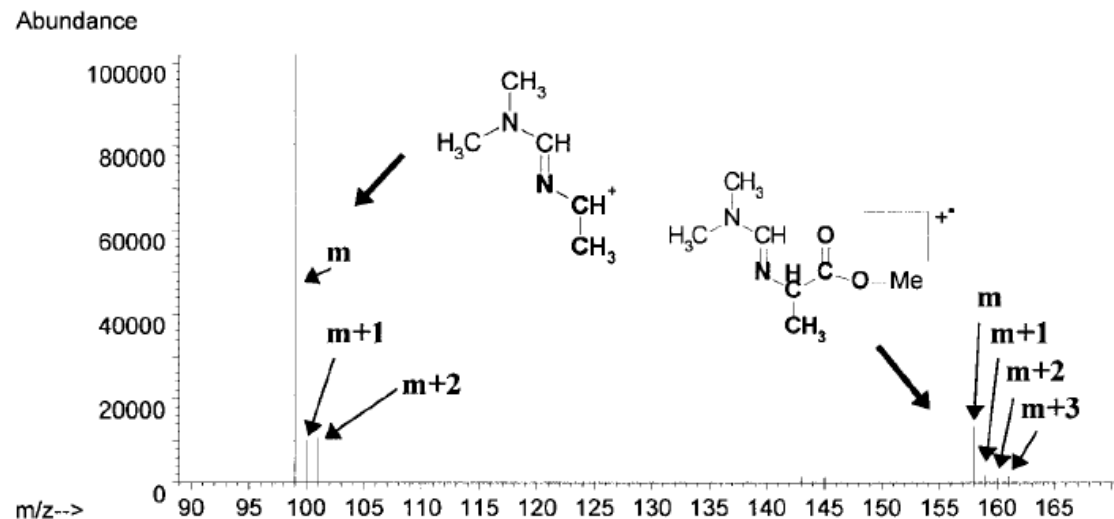
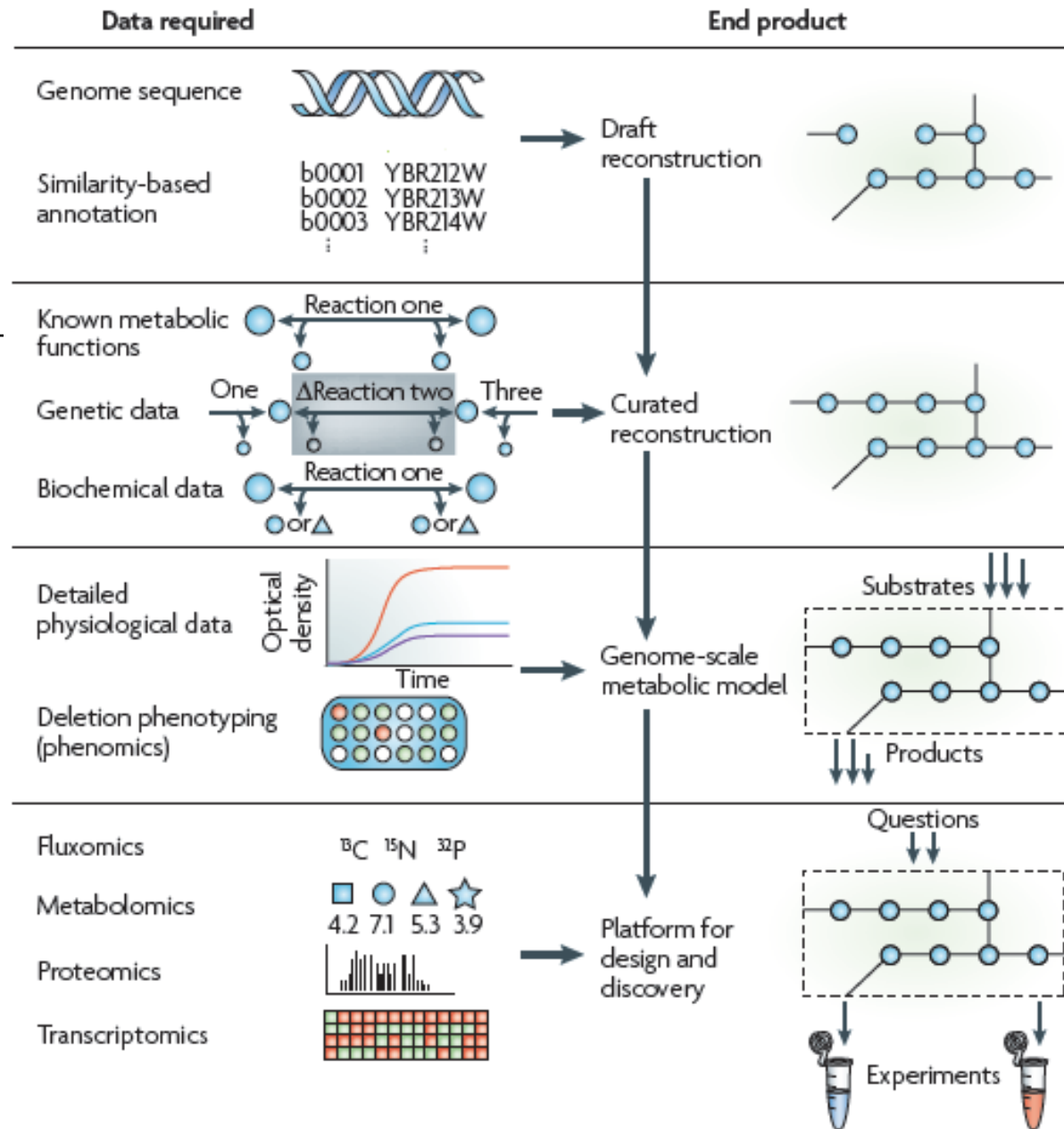


FIG. 1. The eight different positional isotopomers of a C_3 mole. The isotopomers are enumerated from zero to seven, corresponding to binary number formed using the white circles (^{12}C) as zeros and the black circles (^{13}C) as ones.



Redes metabólicas em escala genômica.



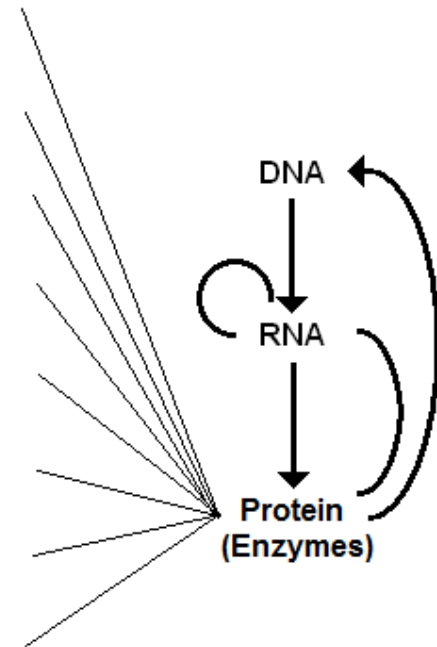
Metabolic Engineering

$$\begin{pmatrix} -1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & -1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & -1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & -1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & -1 & 0 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 & 0 & 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & -1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & -1 & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

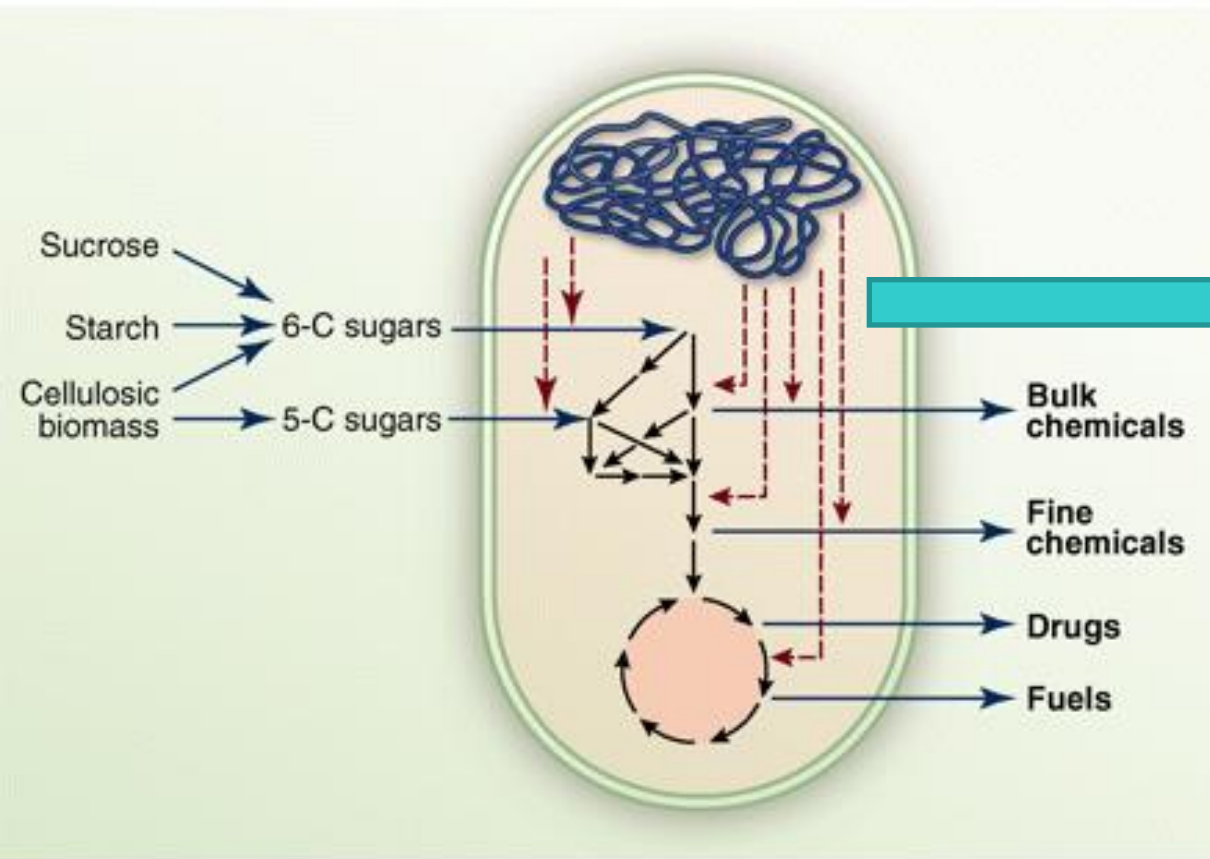
Regulação
metabólica

X

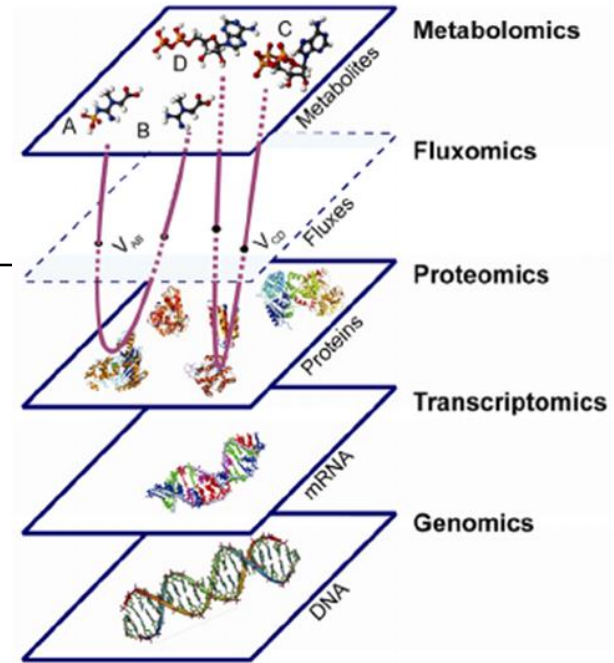
$$\begin{pmatrix} r_1 \\ r_2 \\ r_3 \\ r_4 \\ r_5 \\ r_6 \\ r_7 \\ r_8 \\ r_9 \end{pmatrix}$$



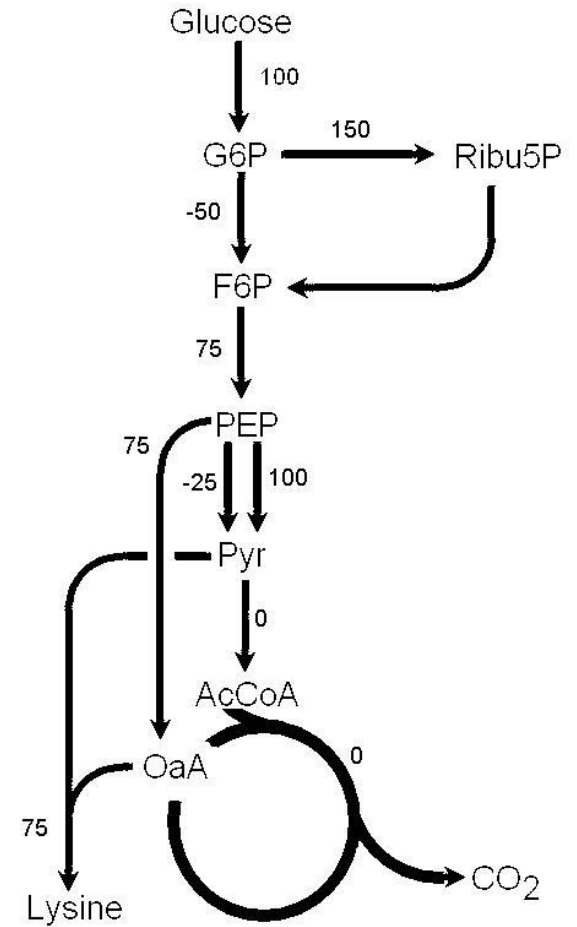
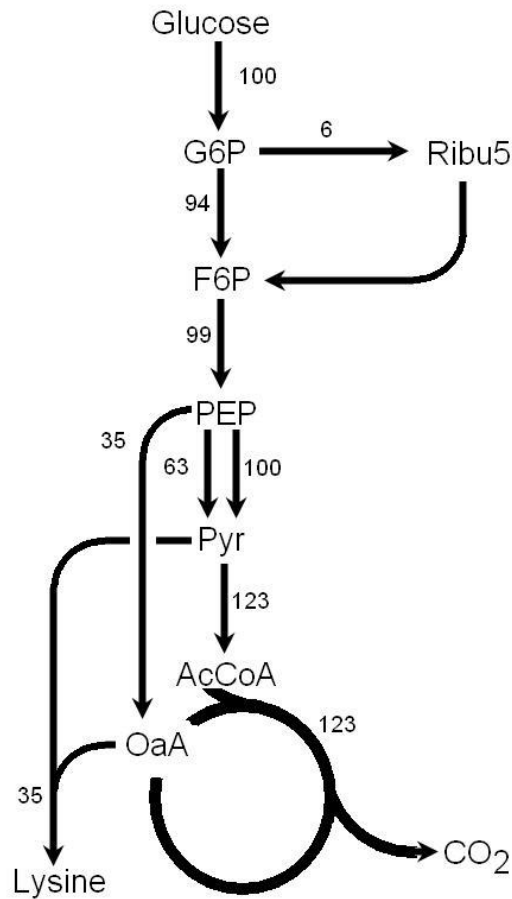
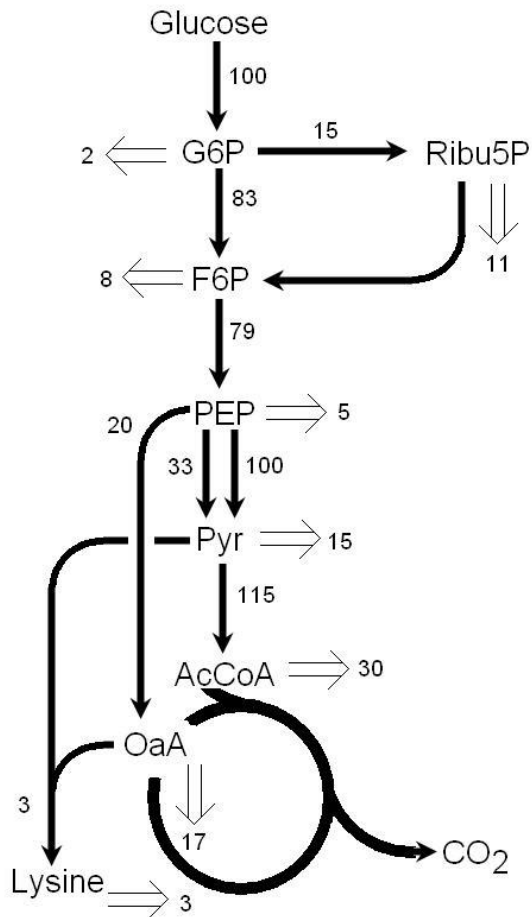
Regulação no
nível hierárquico



Omics revolution

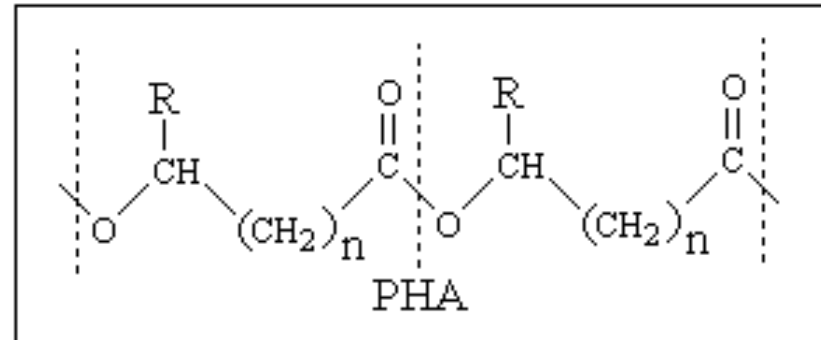
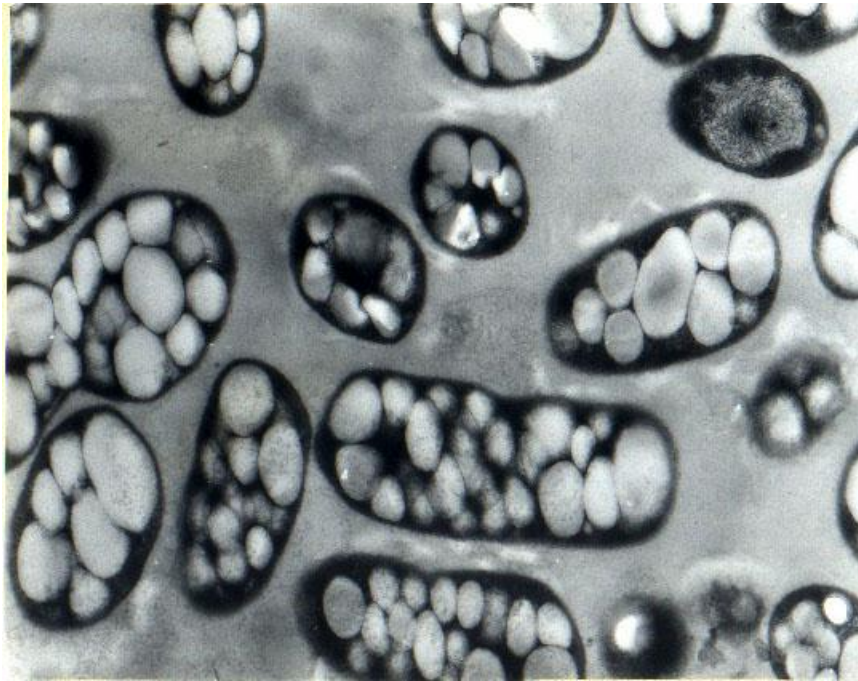


Fluxos metabólicos



Polyhydroxyalkanoates (PHA)

A family of polyesters accumulated by bacteria.



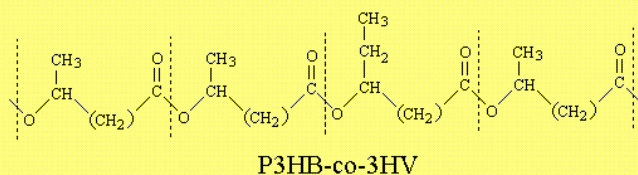
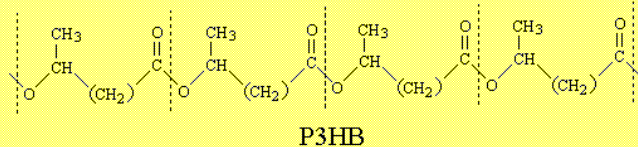
PHA production integrated to a sugar and ethanol mill.



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COPERSUCAR

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R. V. Nonato · P. E. Mantelatto · C. E. V. Rossell

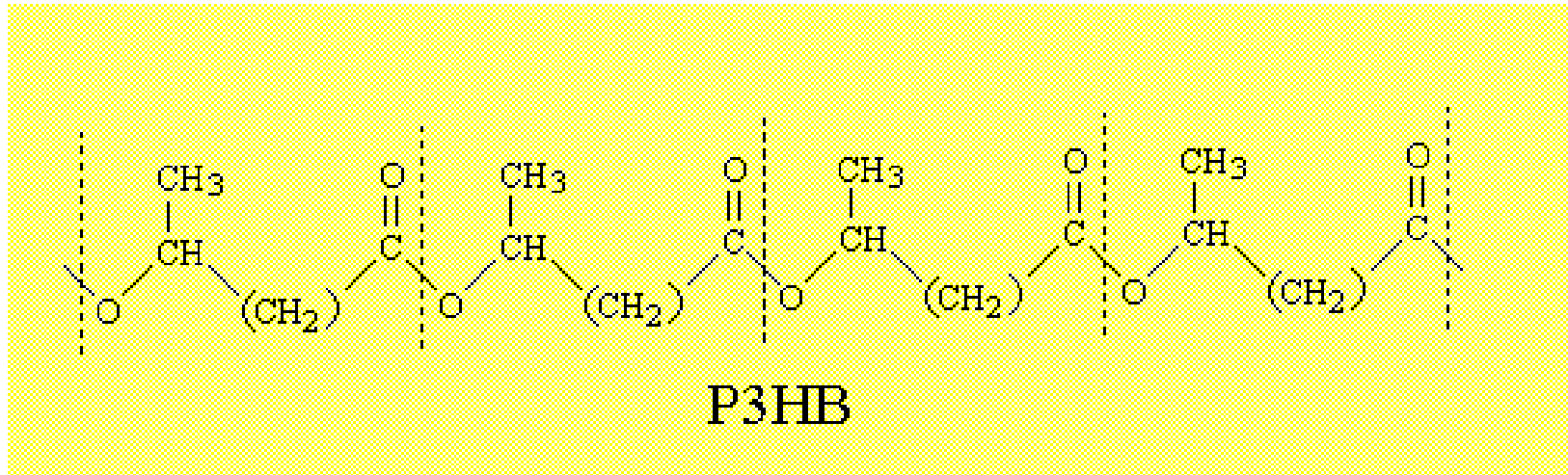
Integrated production of biodegradable plastic, sugar and ethanol

Appl Microbiol Biotechnol (2001) 57:1–5
DOI 10.1007/s002530100732

MINI-REVIEW

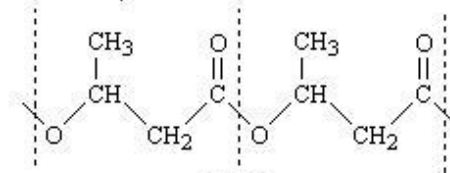
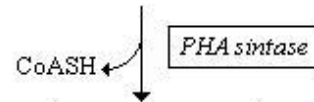
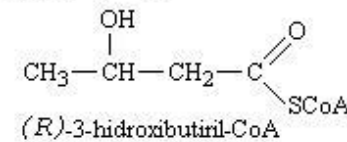
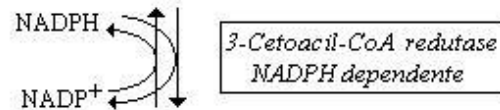
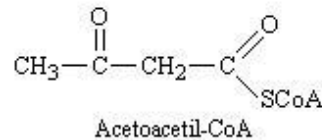
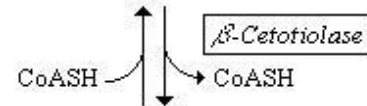
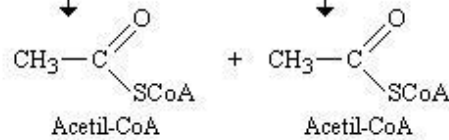
A green cycle for simultaneous poly 3-hydroxybutyric acid, sugar and ethanol production

Polihidroxicanoatos (PHA) Termoplásticos



Carboidratos

Vias de degradação de carboidratos



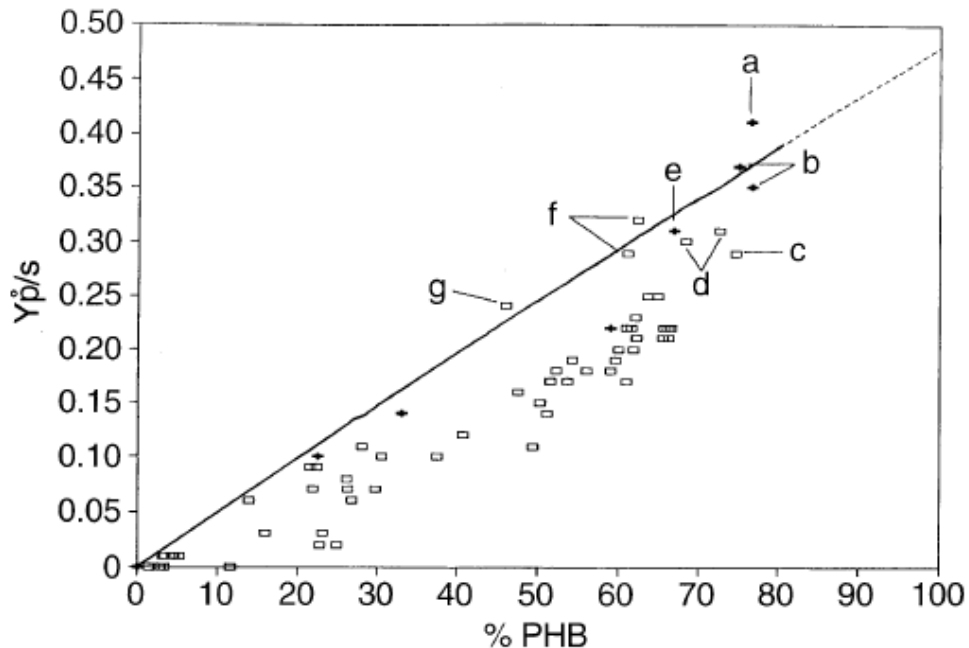
P3HB



P3HB - Produção

Linhagens	Carboidrato	Carboidratos residuais			MSC (g/L)	P3HB	
		S	G	F		%MSC	Y _{P/S}
<i>R.eutropha</i>	G+F	-	1,9	0,0	6,56	76,7	0,35
IPT 086	G+F	-	89,3	0,0	4,32	62,5	0,32
	S	107,4	0,0	0,0	0,35	1,5	0,00
IPT 083	G+F	-	0,0	0,0	6,78	68,5	0,30
	S	0,0	0,0	0,0	6,37	60,9	0,27
IPT 101	G+F	-	0,0	0,0	5,59	74,6	0,29
	S	0,0	0,0	0,0	6,14	68,4	0,29
IPT 076	G+F	-	0,0	0,0	6,31	63,6	0,25
	S	0,0	0,0	0,0	6,46	56,2	0,26

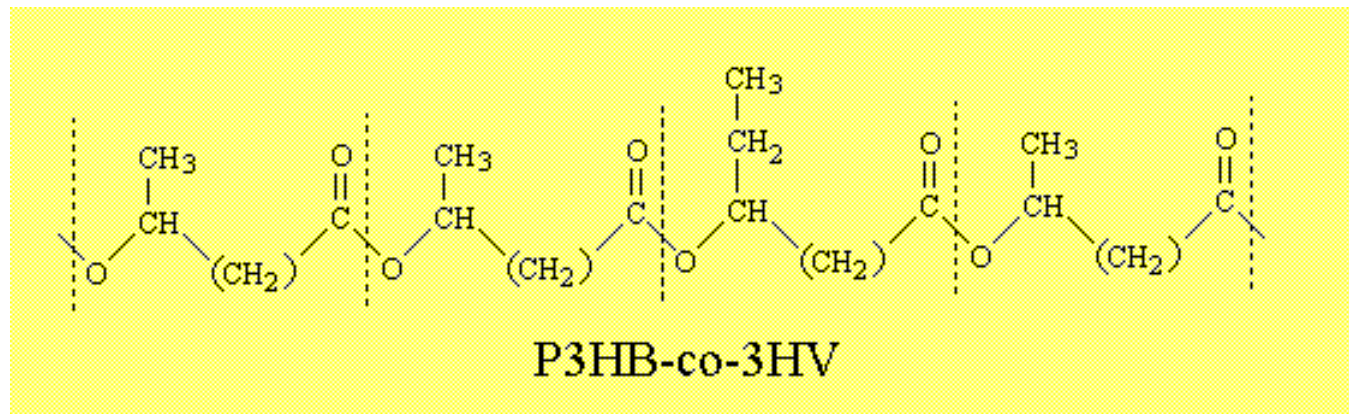
P3HB production from sugarcane carbohydrates



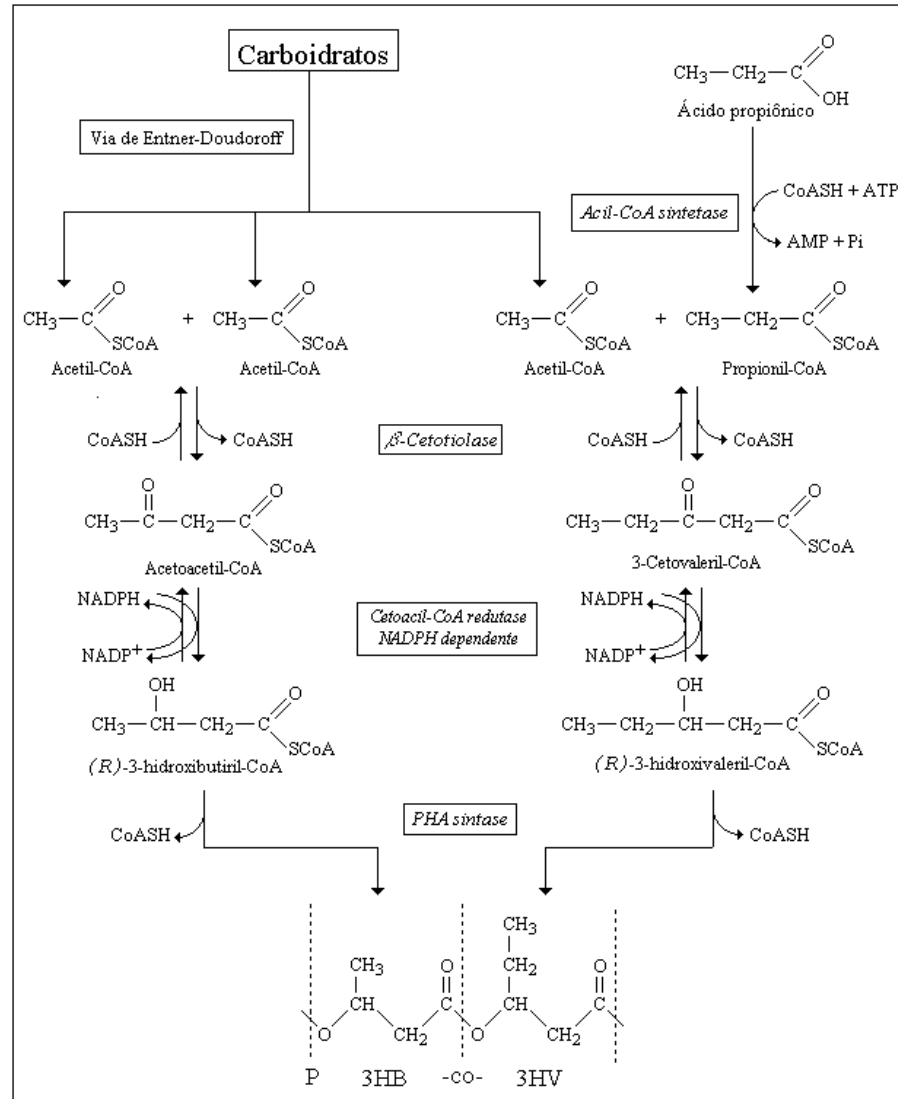
$$Y_{P/C}^O = \frac{\text{PHB}}{\text{PHB} \left(\frac{1}{Y_{P/C}^T} - \frac{1}{Y_{X/C}} \right) + \frac{100}{Y_{X/C}}}$$

Fig. 1 Relation between $Y_{P/C}^O$ and poly-(3-hydroxybutyrate) (PHB) content for different strains isolated from soil (□) or obtained from the culture collection (+) when glucose plus fructose was used as the carbon source. The line represents the values expected when $Y_{P/C}^T = 0.48$ g/g and $Y_{X/C} = 0.50$ g/g. Points related to strains *A. latus* DSM 1123 (a), *A. eutrophus* DSM 545 (b), IPT-101 (c), IPT-083 (d), *A. eutrophus* DSM 428 (e), IPT-086 (f), and IPT-055 (g) are indicated

Análise de fluxos - Balanços metabólicos



Análise de fluxos - Balanços metabólicos



P3HB-co-3HV production from carbohydrates and propionic acid

Strains	CDW (g/l)	Residual carbohydrates (%)	PHA			
			CDW %	3HB (mol%)	3HV (mol%)	$Y_{3HV/PROP}$ (g/g)
<i>A. eutrophus</i> DSM 545	3.92	0.0	71.4	96.1	3.9	0.13
<i>A. latus</i> DSM 1123	0.95	101.8	14.6	55.0	45.0	0.07
<i>P. cepacia</i> DSM 50181	3.35	1.9	38.4	97.3	2.7	0.04
IPT-040	3.77	1.7	32.3	97.1	2.9	0.05
IPT-044	3.92	1.7	51.1	97.1	2.9	0.07
IPT-045	3.73	0.0	49.4	96.2	3.8	0.08
IPT-048	2.97	0.0	44.3	96.2	3.8	0.06
IPT-055	4.27	72.2	1.5	100.0	0.0	0.00
IPT-056	3.60	31.3	30.9	98.5	1.5	0.02
IPT-076	5.06	1.9	56.8	97.1	2.9	0.10
IPT-083	4.89	5.2	56.8	96.9	3.1	0.10
IPT-086 ^a	2.06	75.4	39.0	89.9	10.1	0.09
IPT-098	5.90	0.0	17.7	94.7	5.3	0.07
IPT-101	2.98	41.8	32.3	95.4	4.6	0.05

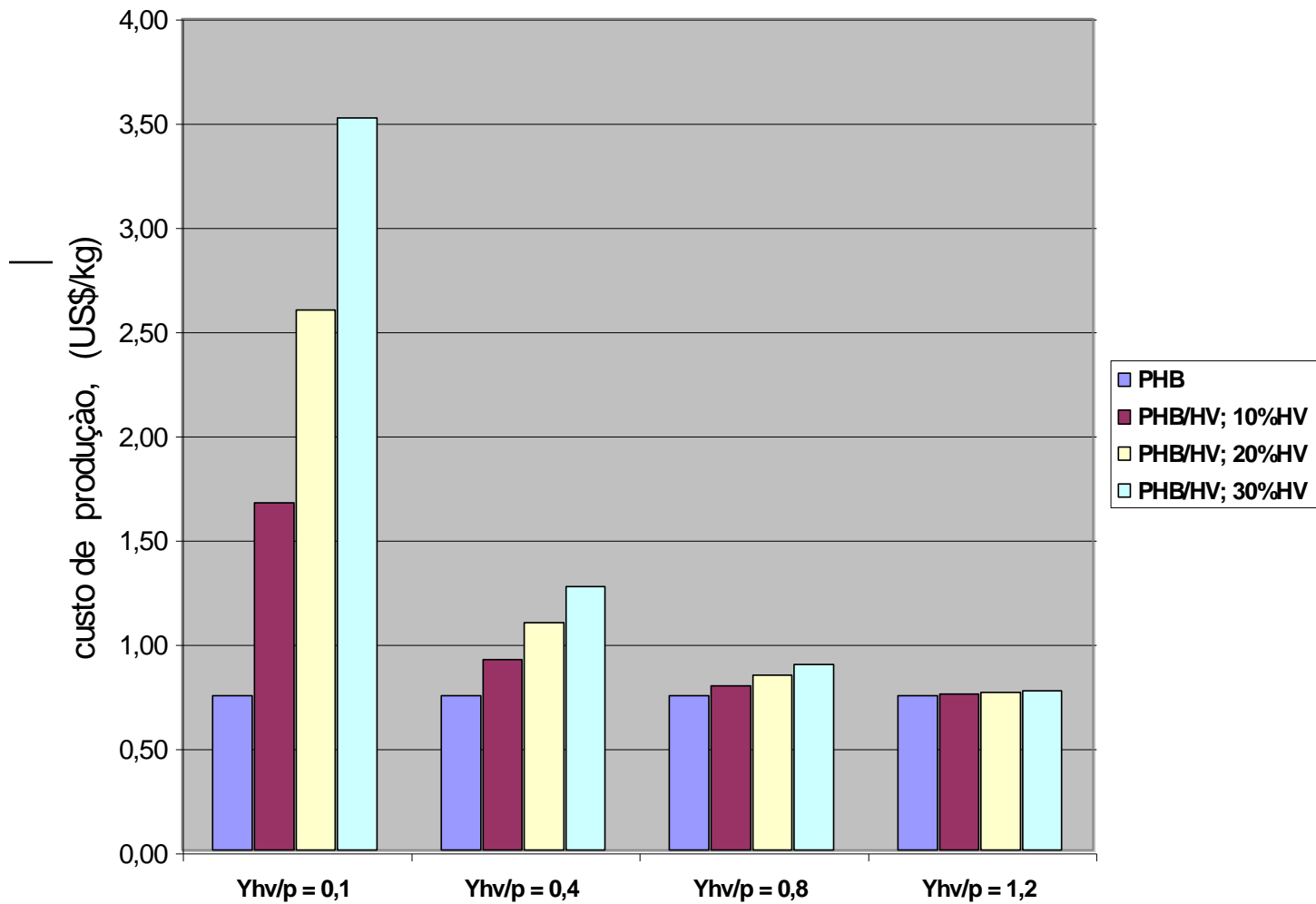
^a Fructose instead of glucose was supplied

Maximum theoretical yield = 1.35 g/g

P3HB-co-3HV production from carbohydrates and propionic acid

Efficiency of *B. sacchari* mutants in converting propionic acid to 3HV units.

Strain	Phenotype	Results			
		3HB mol%	3HV mol%	$Y_{3HV/prp}$ (g/g) ^c	
IPT 101 ^d	wild type	93.8	6.2	0.10	
IPT 183	I	84.1	15.9	0.34	
IPT 185	II	82.6	17.4	0.35	
IPT 190	III	80.1	19.9	0.37	
IPT-195	IV	39.0	61.0	0.81	
IPT 196	IV	33.2	66.8	0.78	
IPT 189	IV	44.7	55.3	0.81	Silva <i>et al.</i> , 2000
IPT 189	feeding strategies of suc/prp			1.34	Rocha <i>et al.</i> , 2008



Carbon source cost

Alta densidade celular

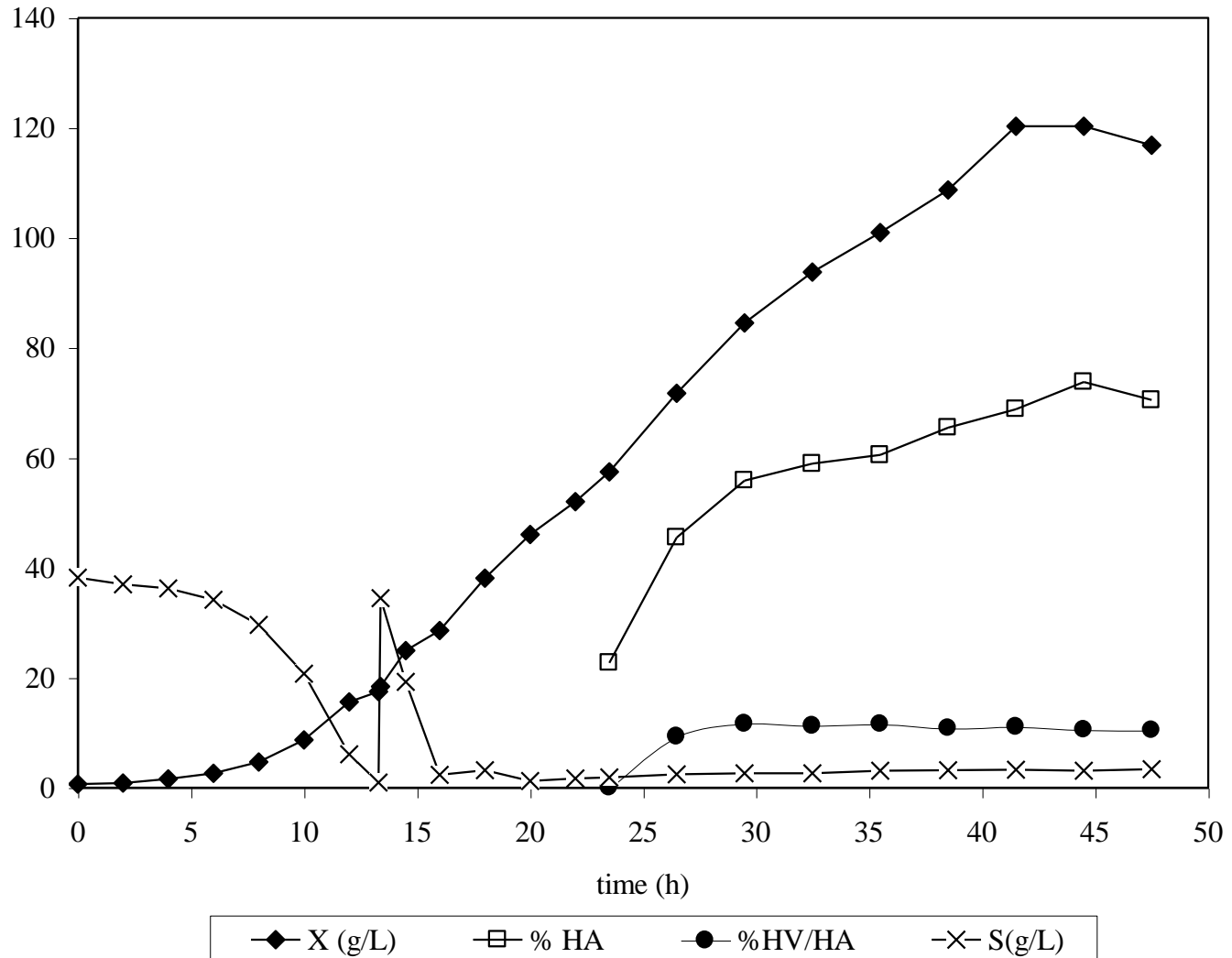


Requirida para aumentar produtividades volumétricas.

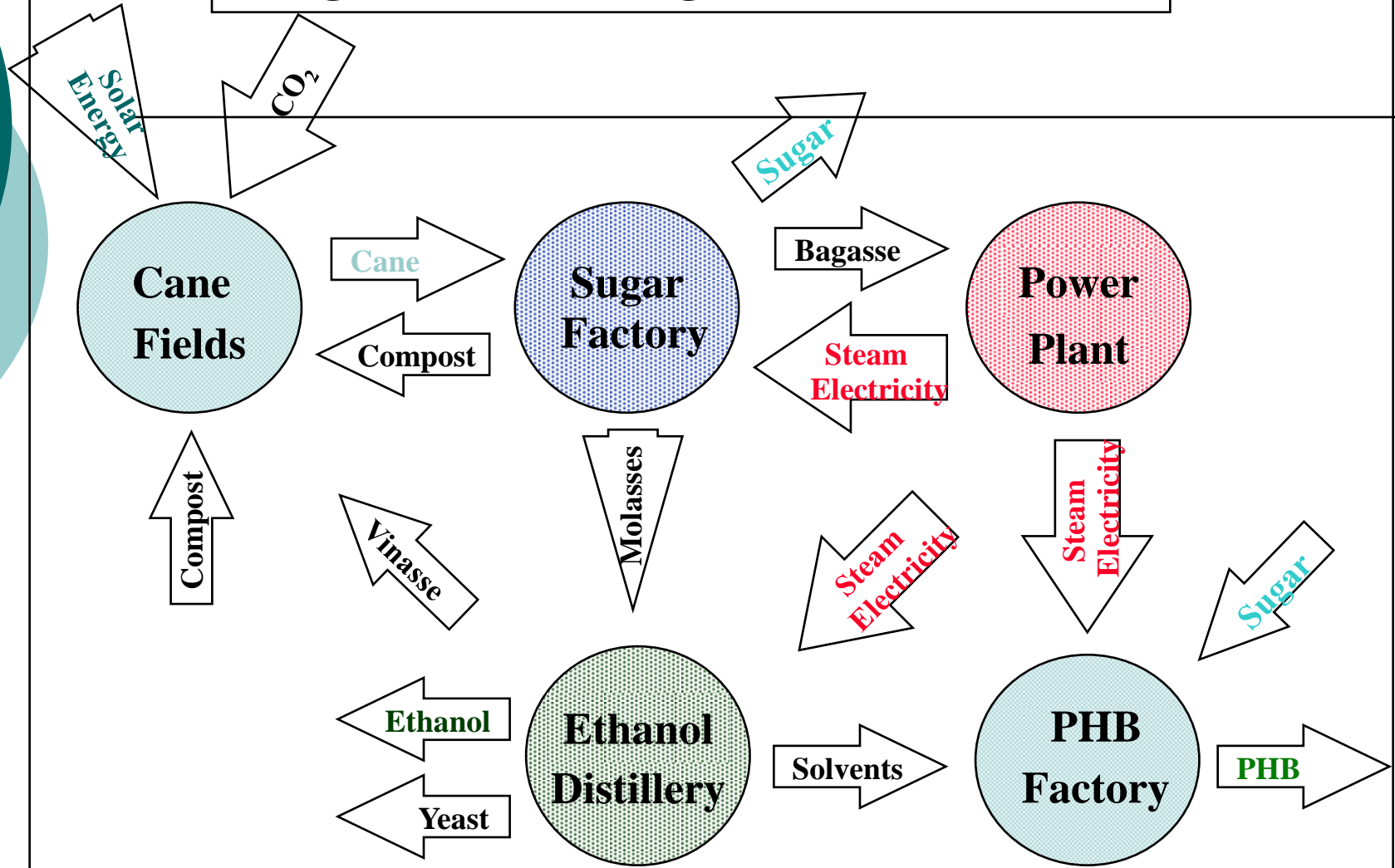
Em processo em batelada alimentada.

Em processo contínuo.

Alta densidade celular



Integrated PHB, Sugar and Ethanol Mill



Produtos de Base biológica

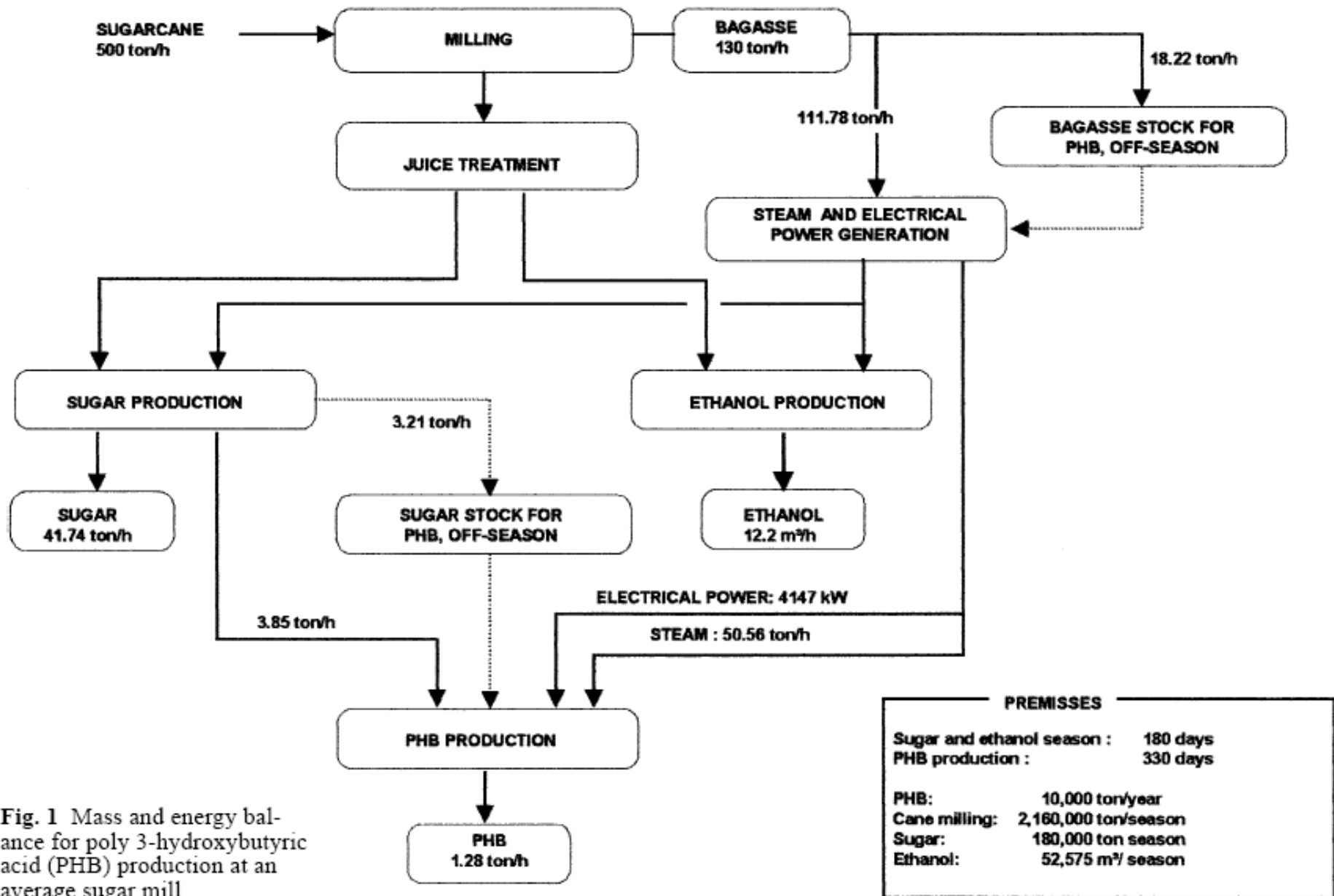


Fig. 1 Mass and energy balance for poly 3-hydroxybutyric acid (PHB) production at an average sugar mill

PREMISES	
Sugar and ethanol season :	180 days
PHB production :	330 days
PHB:	10,000 ton/year
Cane milling:	2,160,000 ton/season
Sugar:	180,000 ton season
Ethanol:	52,575 m³/ season

Integrated production of biodegradable plastic, sugar and ethanol

A green cycle for simultaneous poly 3-hydroxybutyric acid, sugar and ethanol production

Appl Microbiol Biotechnol (2001) 57:1–5
DOI 10.1007/s002530100732

PHA are biobased polymers for biorefineries.

MINI-REVIEW

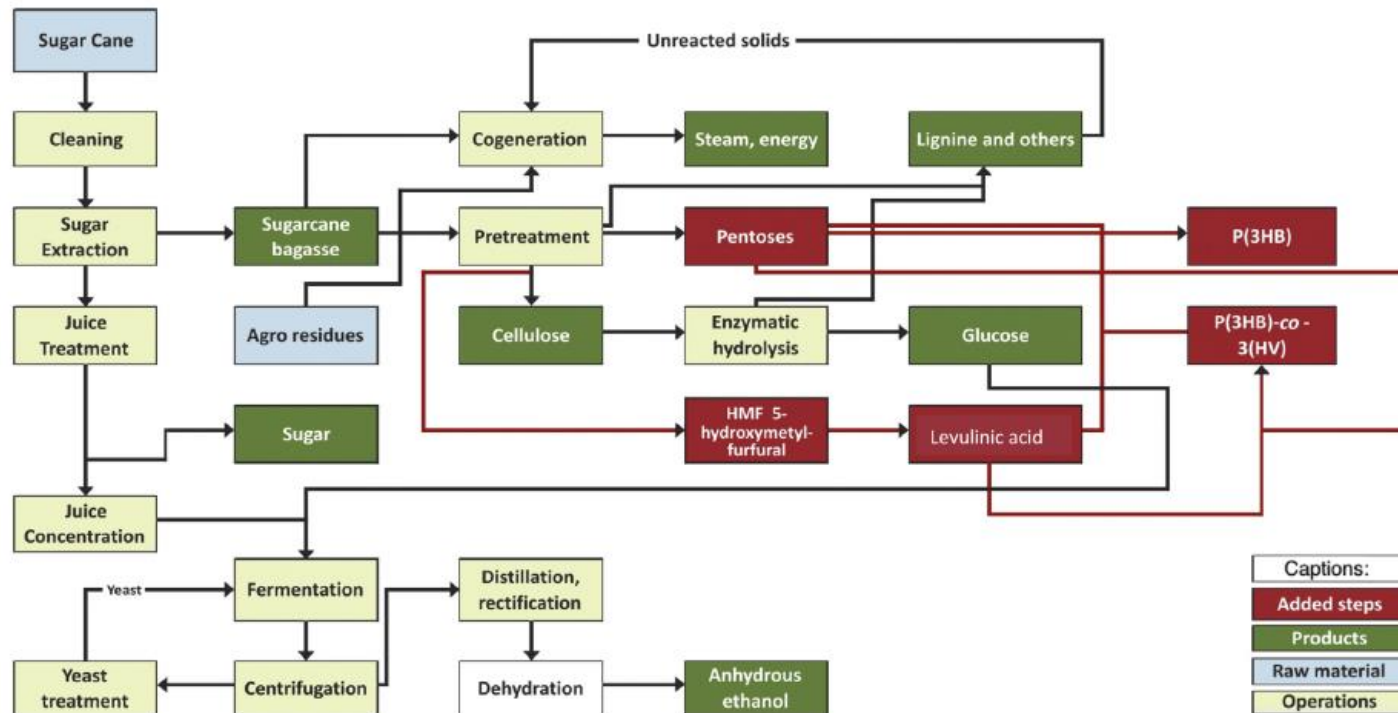


Fig. 1. Steps integrating ethanol and biopolymer production from sugarcane in the context of a biorefinery (modified from Kam & Kam, 2006 [84] by Raicher [82]). Sugar is extracted from the sugarcane, leaving the concentrated juice and bagasse. The concentrated juice is subjected to yeast fermentation to ethanol. Yeast is recycled to a new fermentation. The bagasse and other agricultural residues, such as sugarcane leaves, can have two destinations: cogeneration of steam and energy for the plant, or pretreatment to release cellulose, pentoses and hydroxy methylfurfural (HMF). Cellulose can be hydrolyzed enzymatically, releasing glucose that can be used in the production of ethanol 2G. From HMF, levulinic acid can be produced and used, associated to the pentoses, in the production of PHA.

PHA production integrated to a sugar and ethanol mill.



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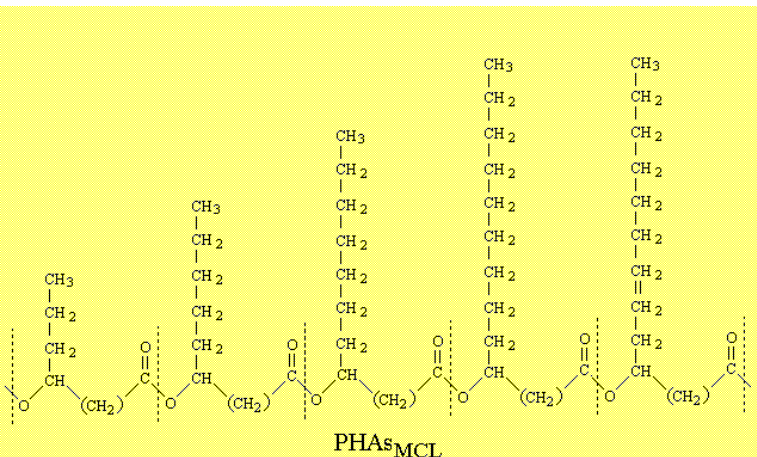
R. V. Nonato · P. E. Mantelatto · C. E. V. Rossell

Integrated production of biodegradable plastic, sugar and ethanol

Appl Microbiol Biotechnol (2001) 57:1–5
DOI 10.1007/s002530100732

MINI-REVIEW

A green cycle for simultaneous poly 3-hydroxybutyric acid, sugar and ethanol production



$$Y_{\text{PHA/G}}^{\text{G}} = \frac{\% \text{PHA}}{\frac{100}{Y_{\text{Xr/G}}} - \frac{\% \text{PHA}}{Y_{\text{Xr/G}}} + \frac{\% 3\text{HHx}}{Y_{3\text{HHx/G}}} + \frac{\% 3\text{HO}}{Y_{3\text{HO/G}}} + \frac{\% 3\text{HD}}{Y_{3\text{HD/G}}} + \frac{\% 3\text{HDd}}{Y_{3\text{HDd/G}}}}$$

Table 1. Production of PHA_{MCL} from carbohydrates by some sugarcane soil isolates.

Bacterial strain	CDW (g/L)	PHA composition (mol%)				PHA (%CDW)	Y _{PHA/G} ^G (g/g)	%Y _{MAX}
		3HHx	3HO	3HD	3HDd			
KT2440	3.96	3.25	12.48	79.88	4.39	48.52	0.127	60.0
LFM046	3.72	5.01	21.85	71.83	1.31	60.51	0.161	62.9
LFM047	2.21	1.72	22.88	61.81	13.58	13.99	0.023	34.3
LFM050	2.50	0.92	15.98	67.75	15.35	18.55	0.037	41.9
LFM065	4.06	0.00	8.88	87.81	3.30	30.52	0.084	60.6

CDW – Cell dry weight

3HHx - 3-hydroxyhexanoic acid

3HO - 3-hydroxyoctanoic acid

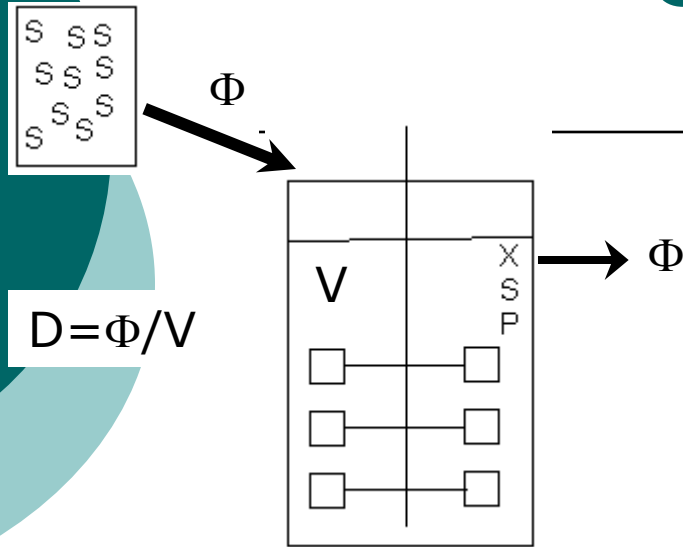
3HD - 3-hydroxydecanoic acid

3HDd - 3-hydroxydodecanoic acid

Y_{PHA/G}^G – global PHA yield from glucose

%Y_{MAX} - percentual of the maximum theoretical yield.

Cultivo contínuo



$$V \frac{dX}{dt} = \Phi X_0 - \Phi X + V \left(\frac{dX}{dt} \right)_{\text{crescimento}}$$

Sendo que

$$\left(\frac{dX}{dt} \right)_{\text{crescimento}} = \mu X \quad D = \frac{\Phi}{V}$$

Temos

$$\frac{dX}{dt} = D (X_0 - X) + \mu X$$

Como $X_0 = 0$ (meio estéril)

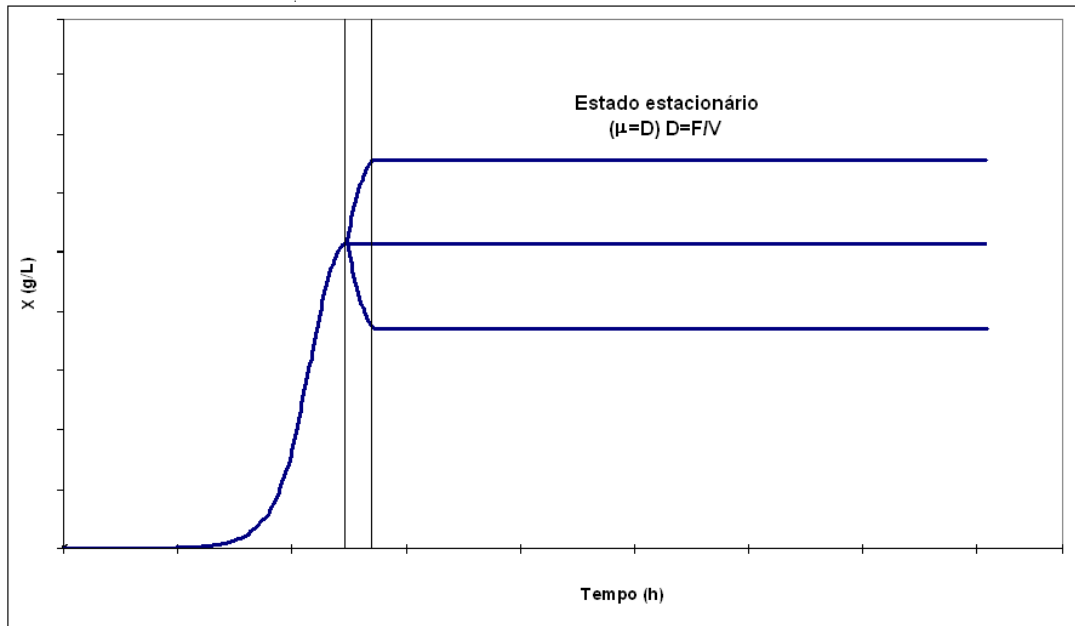
$$\frac{dX}{dt} = \mu X - DX$$

No estado estacionário $dX/dt = 0$

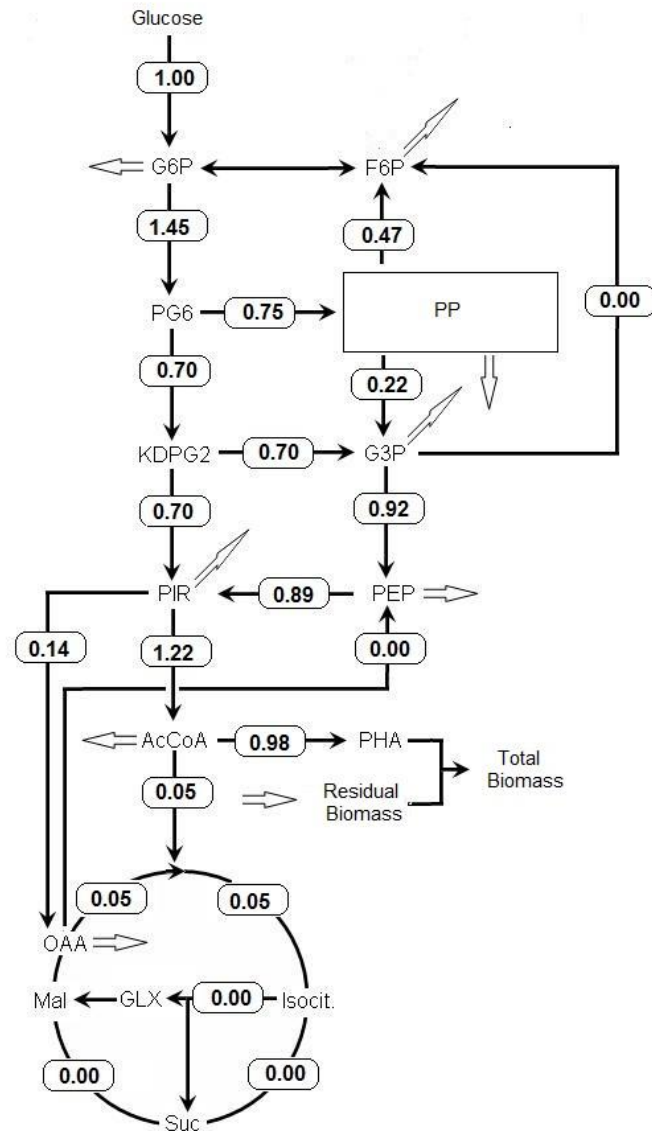
$$\mu X = DX$$

OU

$$\mu = D$$



Elementary Mode analysis



Optimal fluxes distribution for PHA production by *Pseudomonas* sp. from glucose.

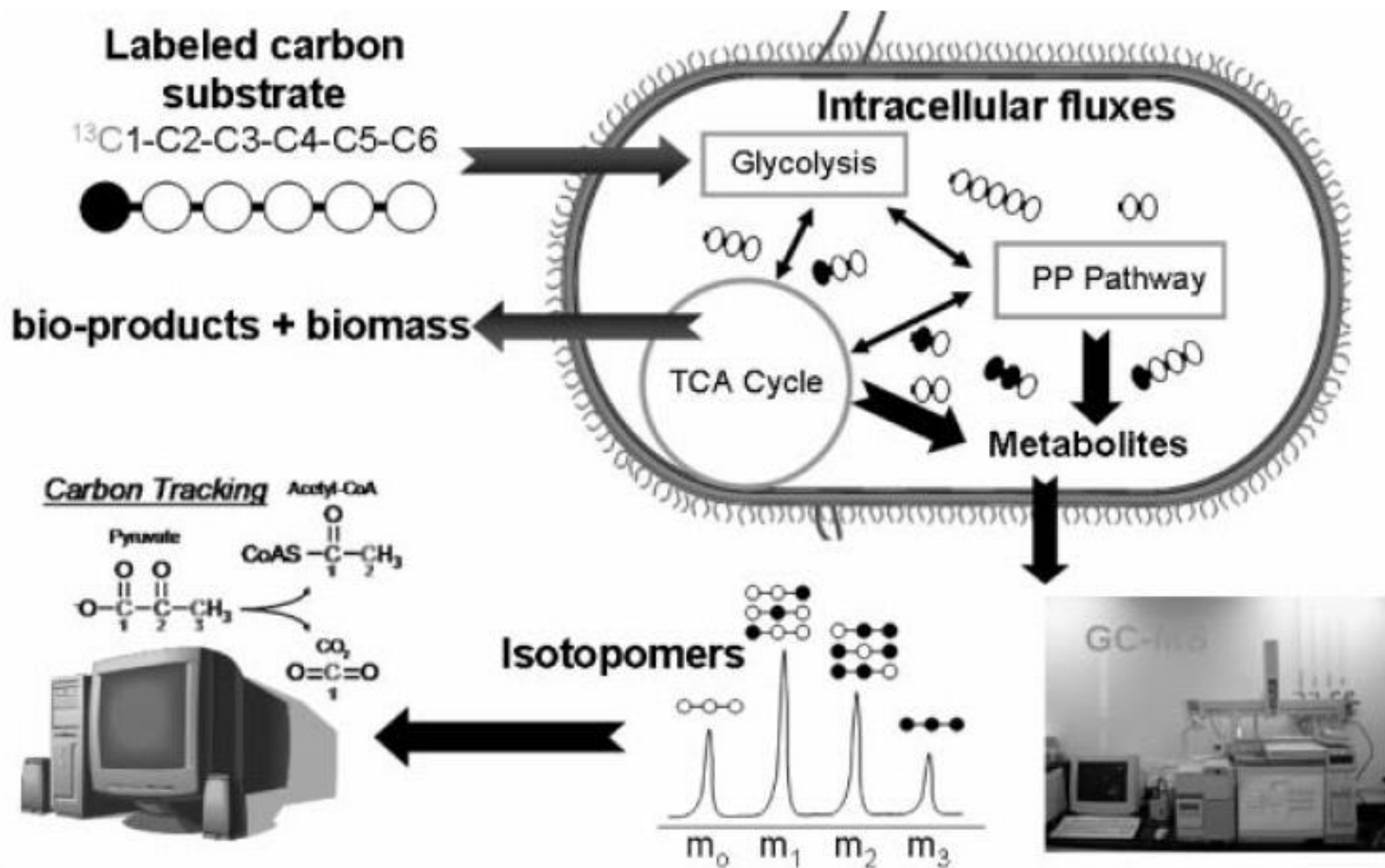
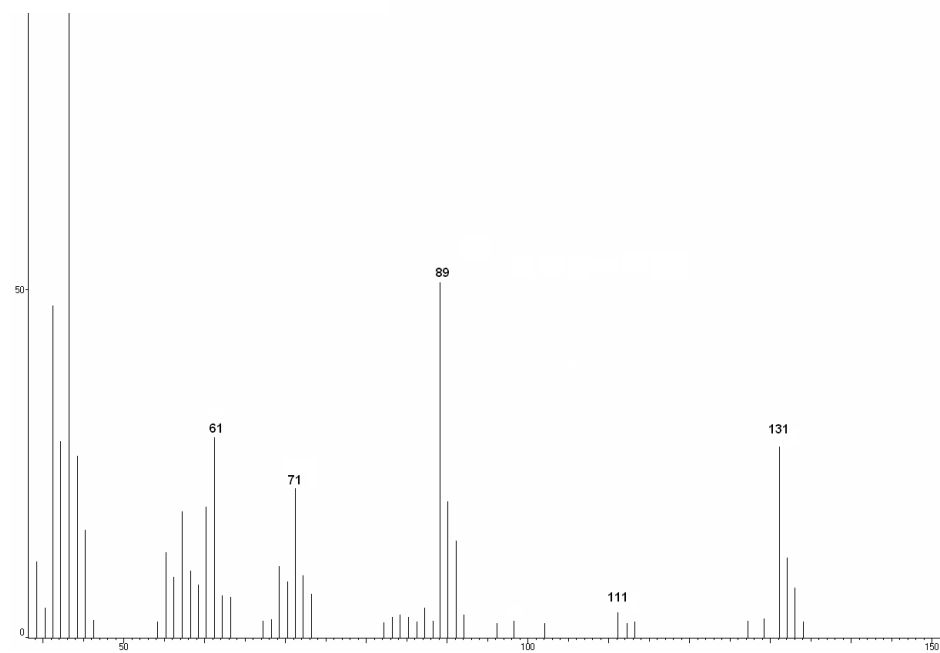
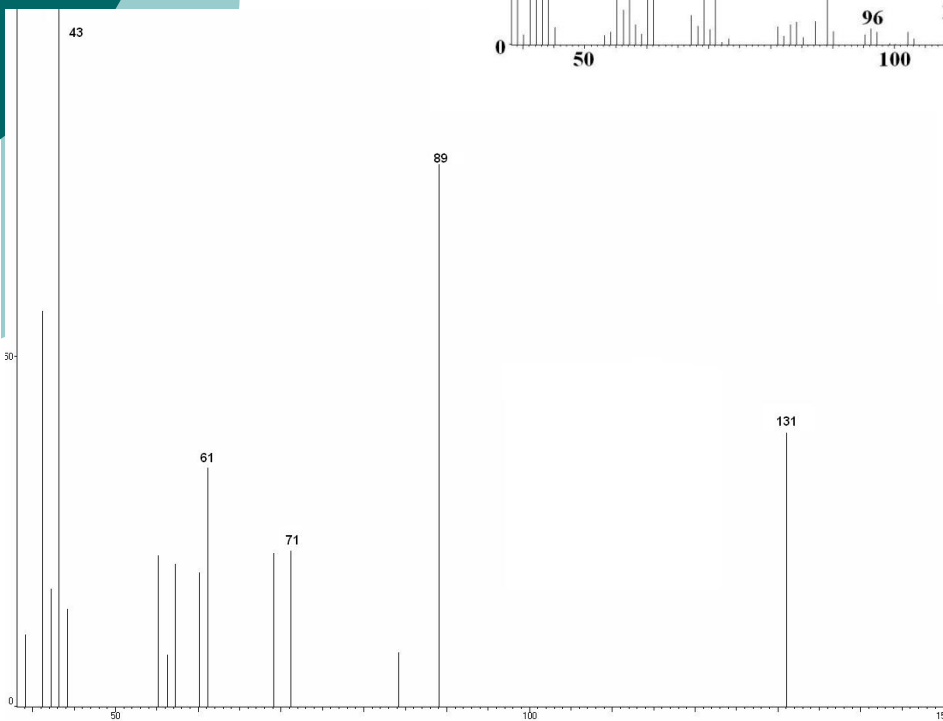
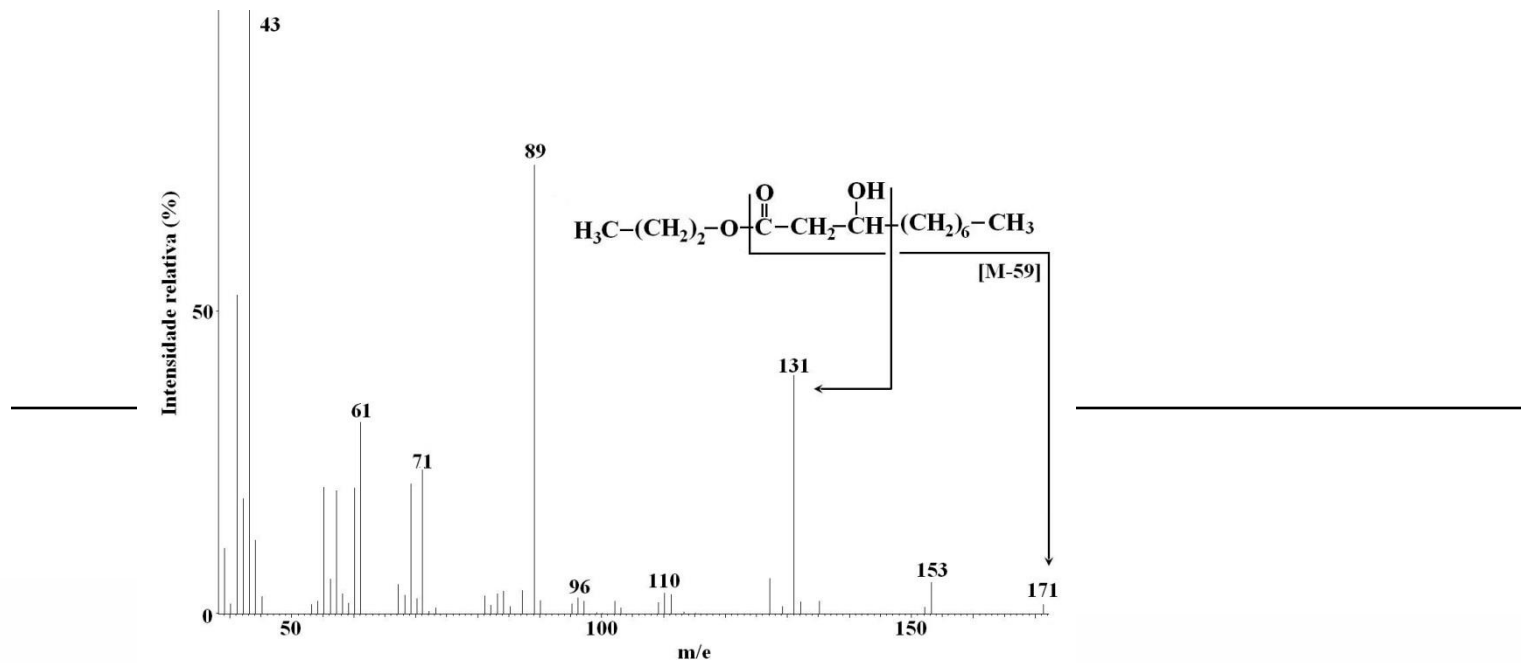
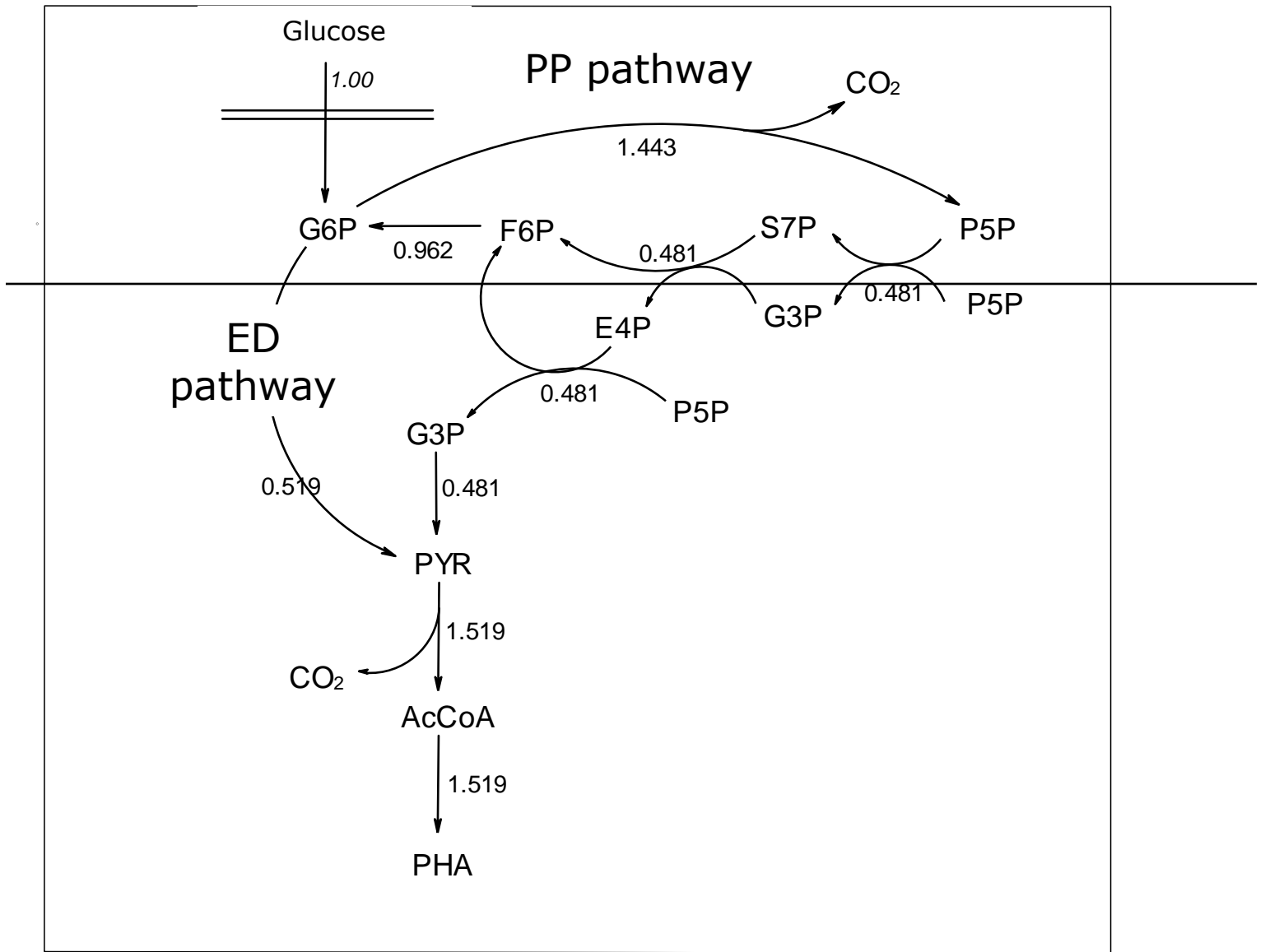


FIGURE 2. Protocol for ^{13}C -based flux analysis. [Color figure can be viewed in the online issue, which is available at www.interscience.wiley.com.]

Análise de fluxos – Experimentos com traçadores





Genomic scale metabolic networks



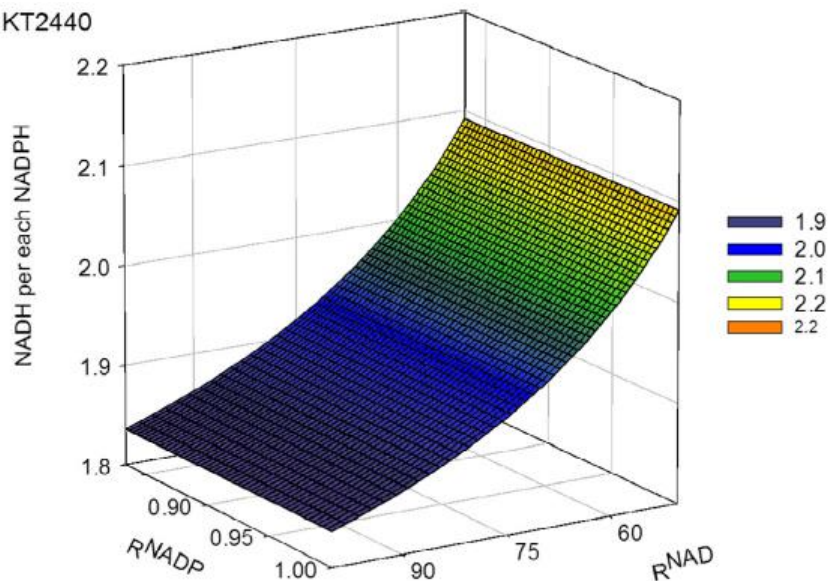
Draft Genome Sequence of *Pseudomonas* sp. Strain LFM046, a Producer of Medium-Chain-Length Polyhydroxyalkanoate

Julliana Cardinali-Rezende,^a Paulo Moises Raduan Alexandrino,^{a,b} Rafael Augusto Theodoro Pereira de Souza Nahat,^a Débora Parrine Vieira Sant'Ana,^{a*} Luizlana Ferrelra Silva,^a José Gregório Cabrera Gomez,^a Marilda Kelco Taciro^a

Institute of Biomedical Sciences, University of São Paulo, São Paulo, Brazil^a; Department of Computer Science, Institute of Mathematics and Statistics, University of São Paulo, São Paulo, Brazil^b

* Present address: Débora Parrine Vieira Sant'Ana, Department of Bioresource Engineering, McGill University, Montreal, Quebec, Canada.

P. putida KT2440



E. coli

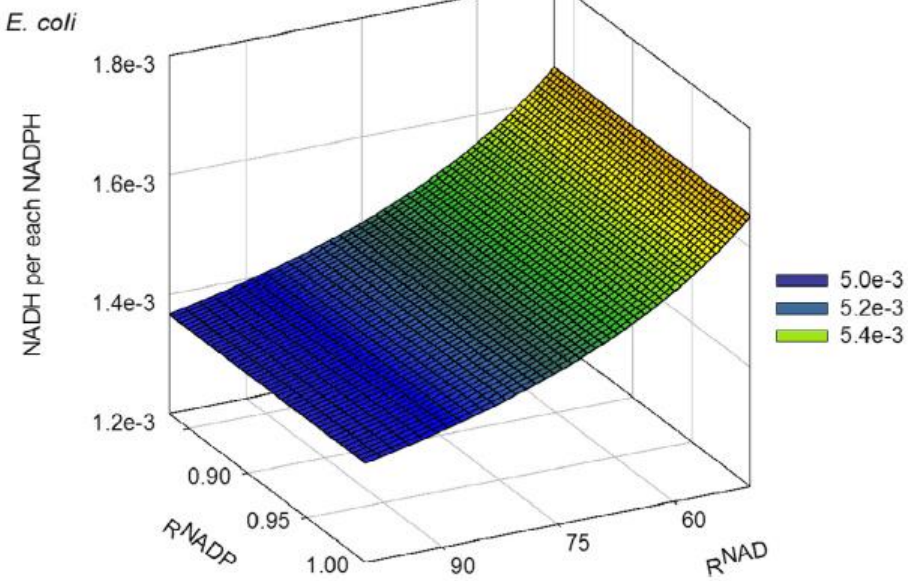


Fig. 4. Relative productions of NADH respect to NADPH during the oxidation of G6P catalyzed by the enzymes PputG6PDH-1 (top) and EcG6PDH (bottom), depending on the concentrations ratios (*R*) of the oxidized and reduced forms of NAD(H) and NADP(H). The represented values were obtained assuming a concentration of G6P of 1 mM.

Table 3

GND specific activities, using NAD or NADP as cofactors, registered in crude cellular extracts from *E. coli* MG1655 and *P. putida* KT2440 grown in mineral medium with glucose as the sole carbon source.

Bacteria	^a Activity with NAD (nmol mg _{prot} ⁻¹ min ⁻¹)	^a Activity with NADP (nmol mg _{prot} ⁻¹ min ⁻¹)
<i>E. coli</i> MG1655	8 ± 1.5	103 ± 0.4
<i>P. putida</i> KT2440	25 ± 0.4	6 ± 0.4



Journal homepage: www.eurpub.com/locatereferedpubenduo



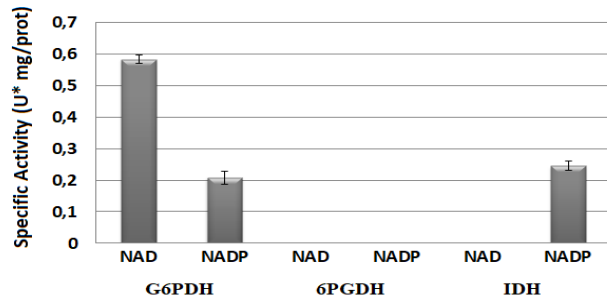
Quantifying NAD(P)H production in the upper Entner–Doudoroff pathway from *Pseudomonas putida* KT2440

Karel Olavarria*, Marina Pupke Marone, Henrique da Costa Oliveira, Juan Camilo Roncallo, Fernanda Nogales da Costa Vasconcelos, Luiziana Ferreira da Silva, José Gregório Cabrera Gomez

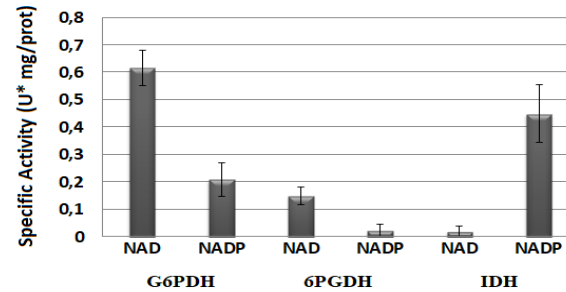
Departamento de Microbiologia, Instituto de Ciências Biomédicas, Universidade de São Paulo, Brazil



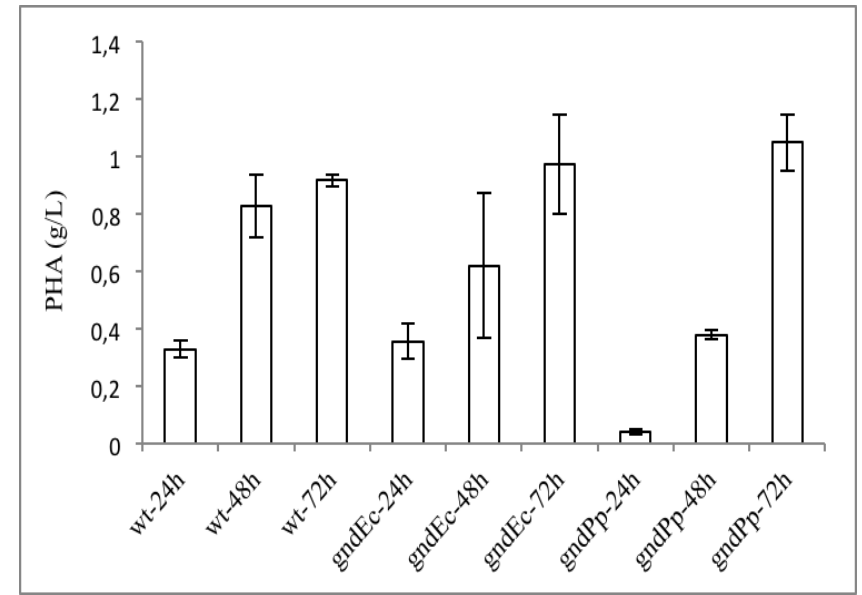
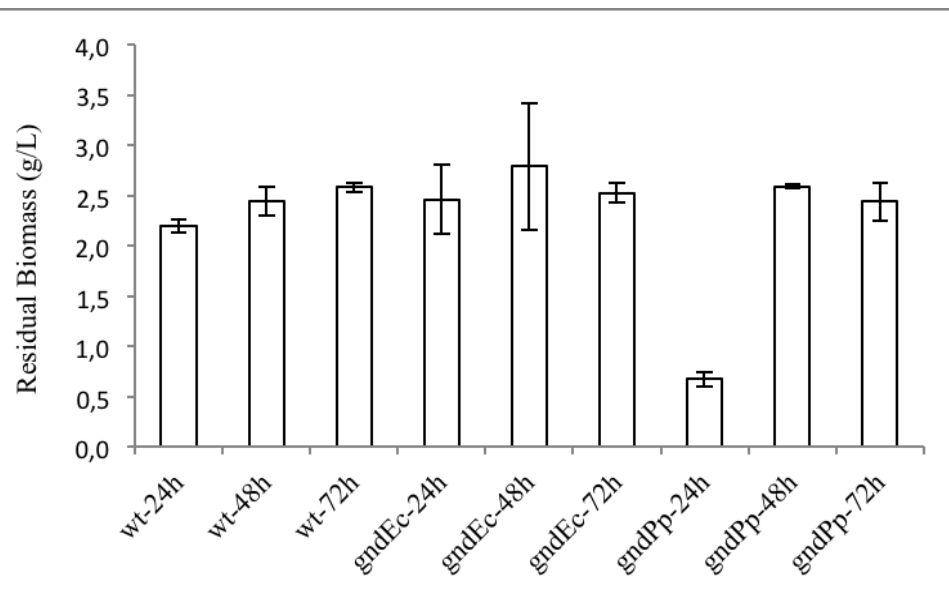
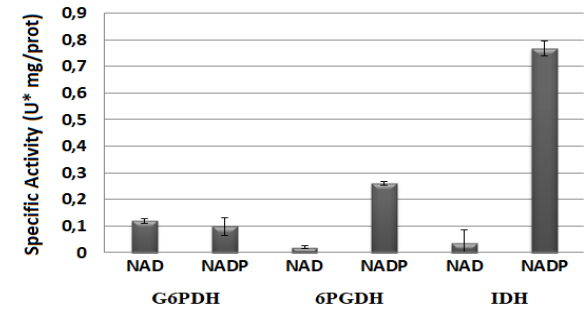
Pseudomonas sp. LFM046



Pseudomonas putida KT2440



Escherichia coli MG1655

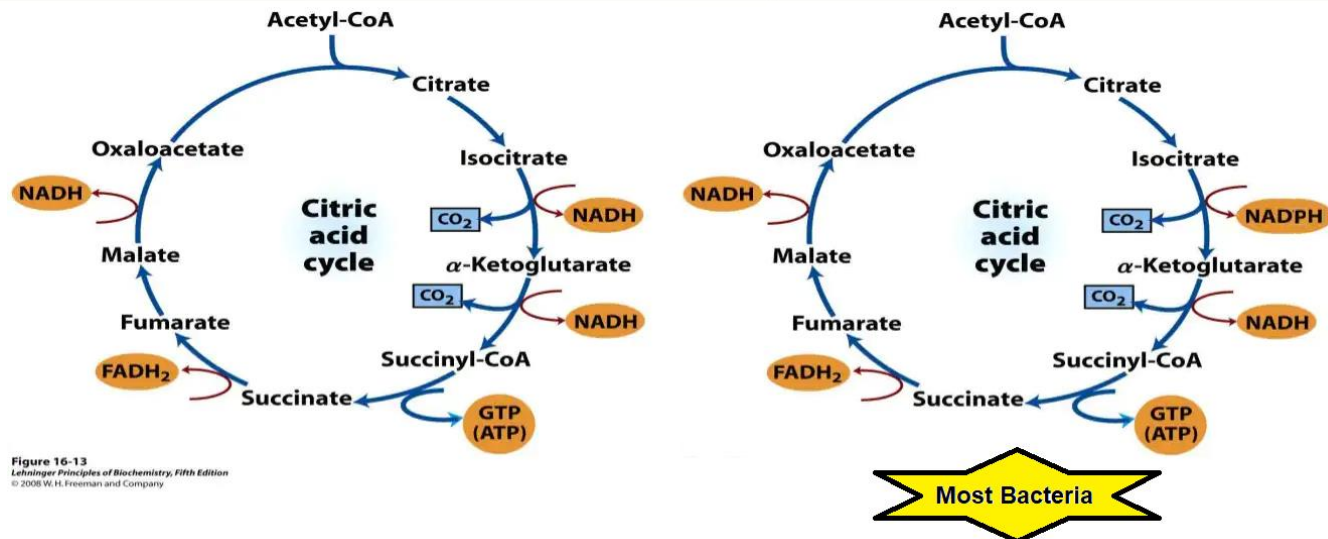
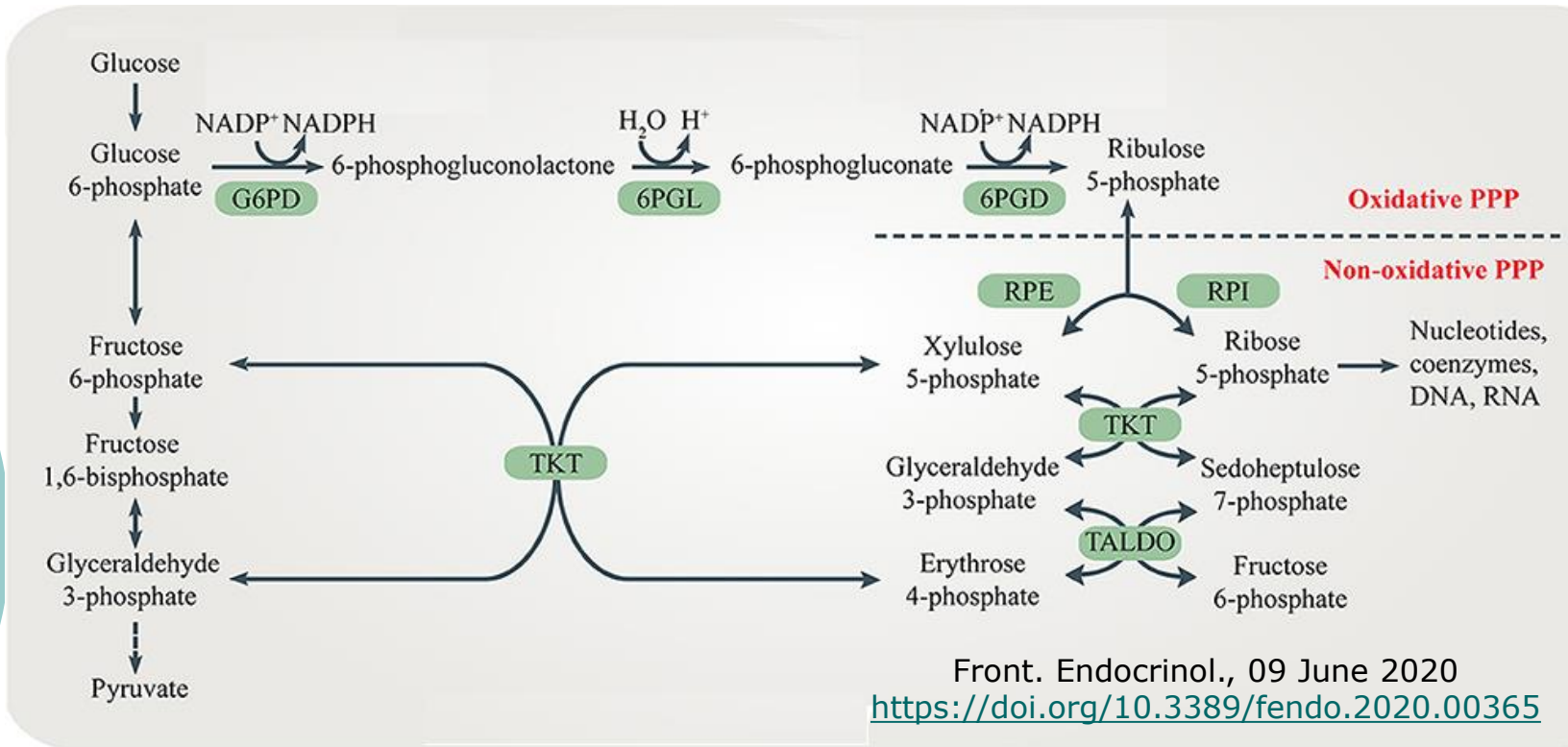


The relevance of enzyme specificity for coenzymes and the presence of 6-phosphogluconate dehydrogenase for polyhydroxyalkanoates production in the metabolism of *Pseudomonas* sp. LFM046

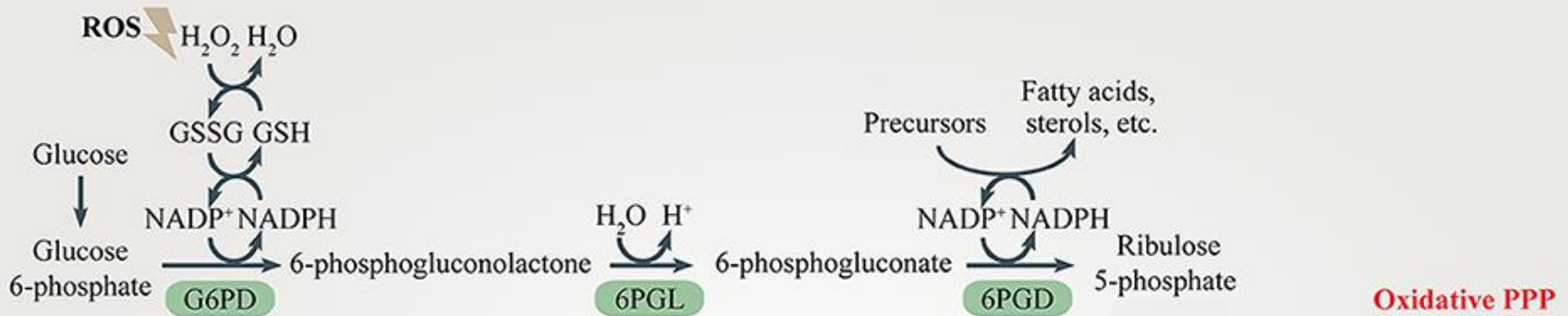
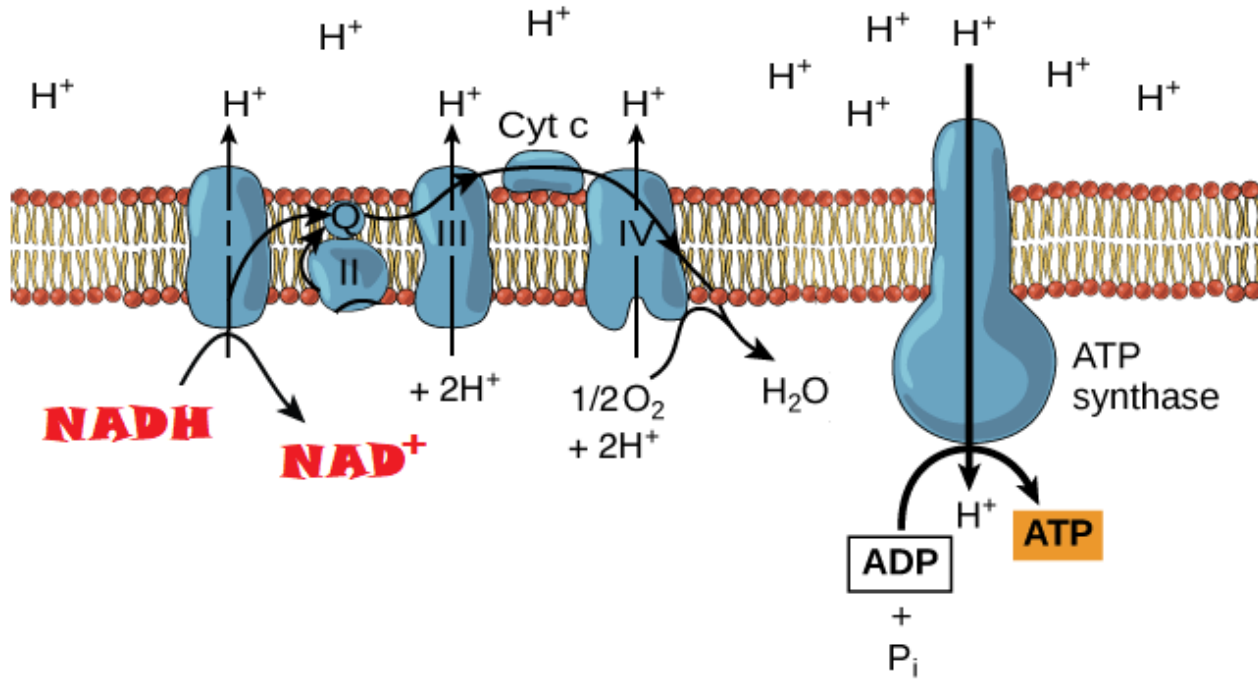
Juliana Cardinali-Rezende ^{a,c,*}, Alex Di Genova ^b, Rafael A.T.P.S. Nahat ^a, Alexander Steinbüchel ^{c,f}, Marie-France Sagot ^b, Rafael S. Costa ^{d,e}, Henrique C. Oliveira ^a, Marilda K. Taciro ^a, Luiziana F. Silva ^a, José Gregório C. Gomez ^{a,**}



Different enzymes specificity: NADH versus NADPH



Different metabolic roles: NADH *versus* NADPH



Front. Endocrinol., 09 June 2020

<https://doi.org/10.3389/fendo.2020.00365>



Imagem de autoria de Nara Guimarães

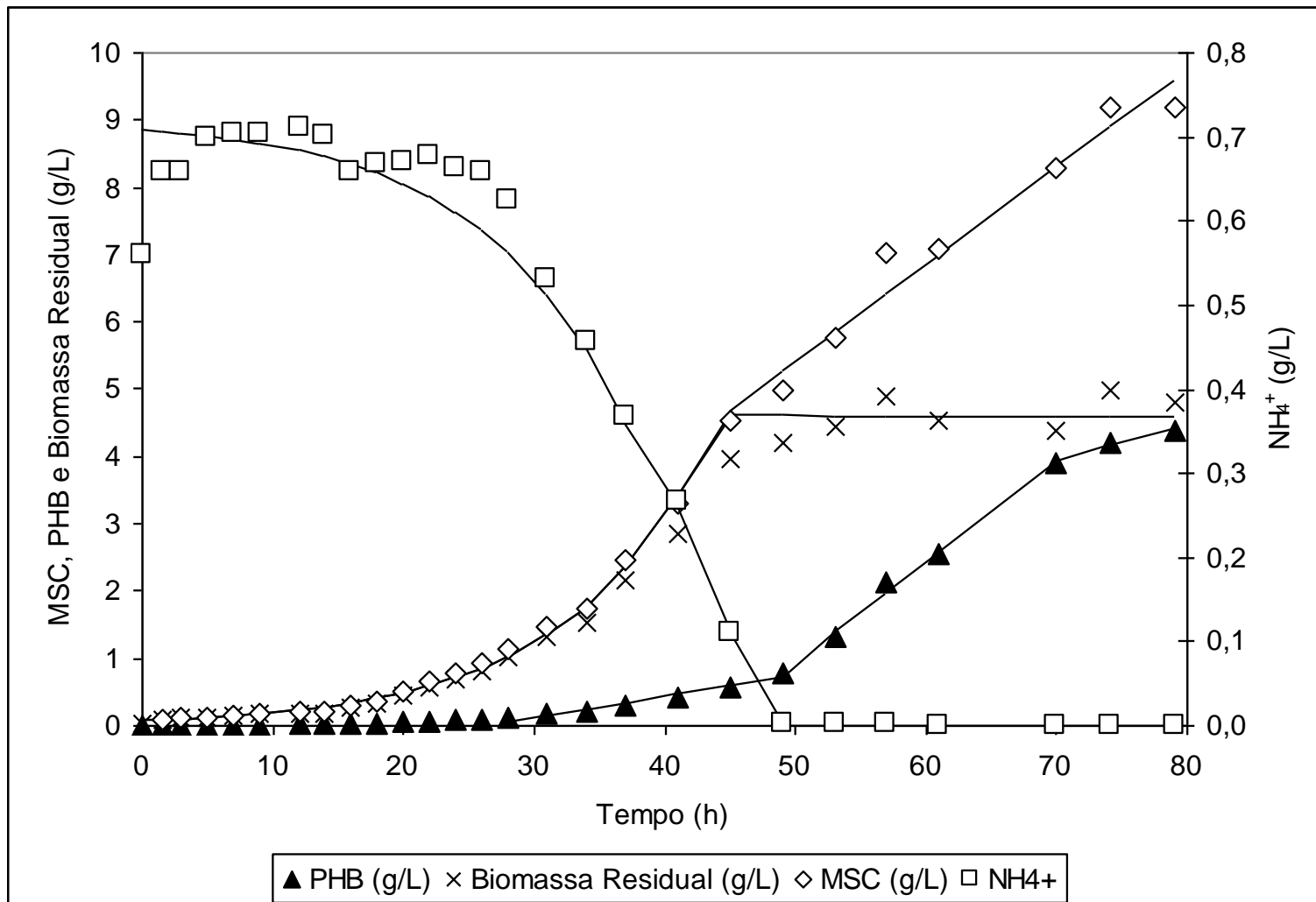
Flux distribution

Best labeling pattern

Genomic data
Enzymes activity

Experimental Data





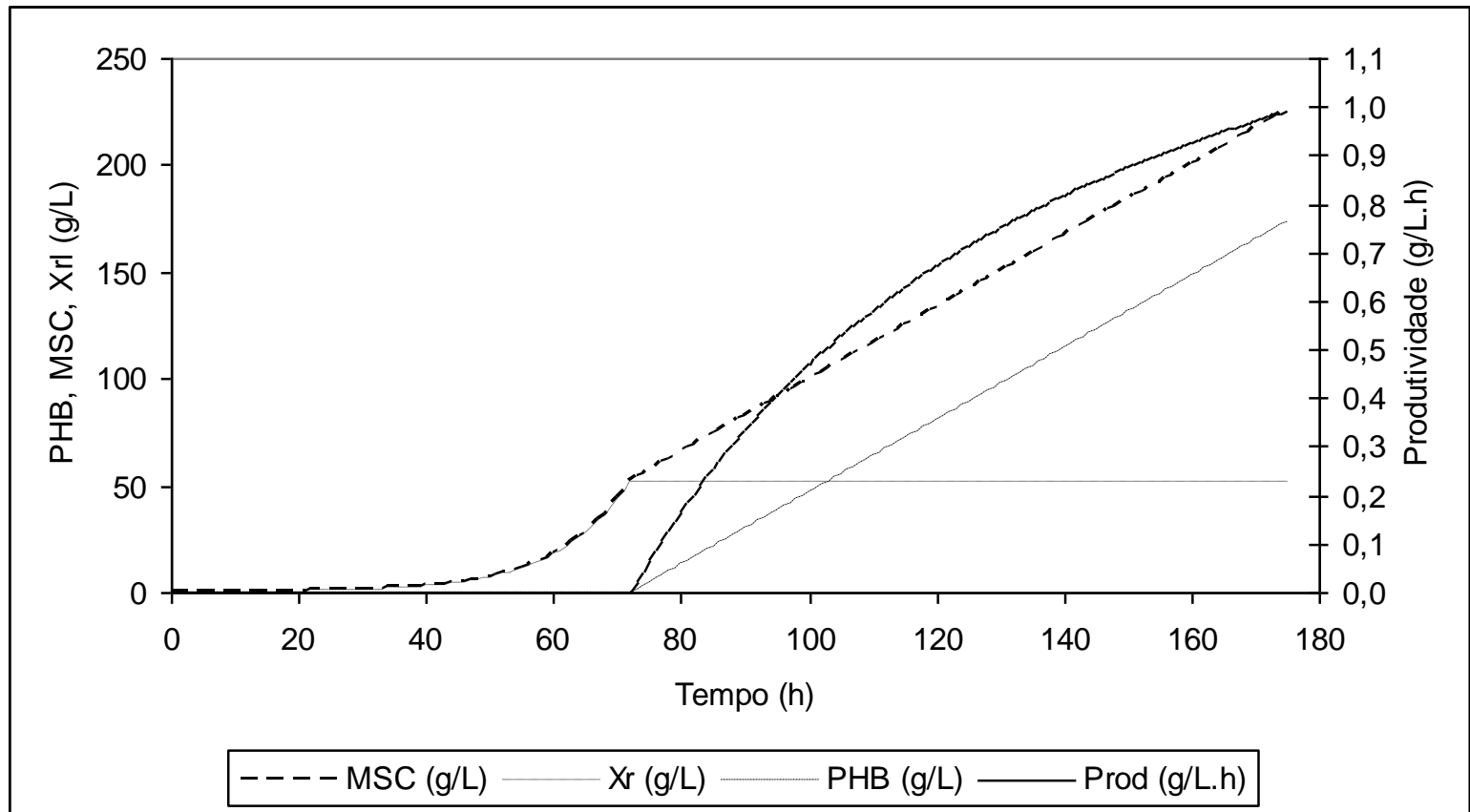
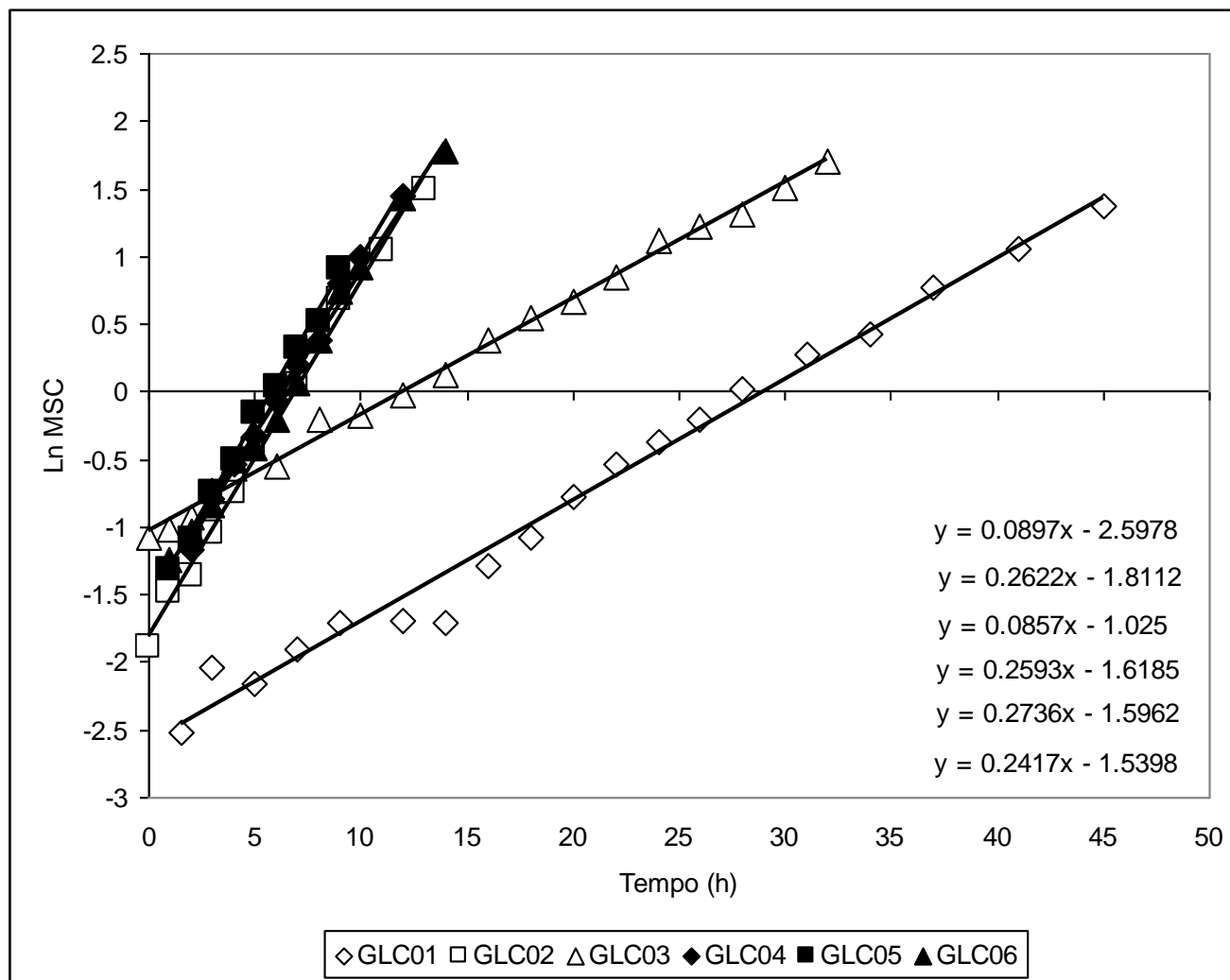
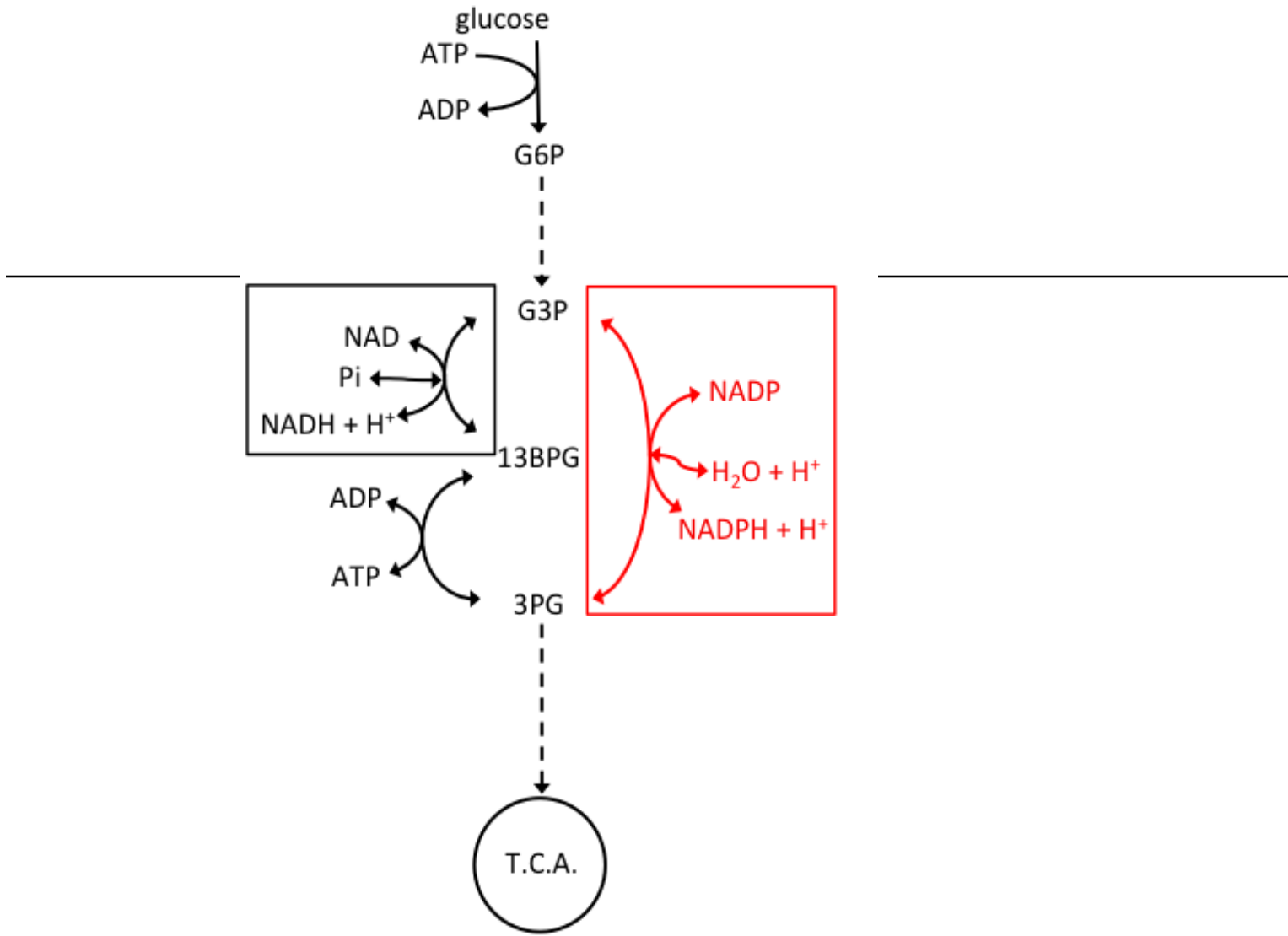


Figura 13. Simulação de processo de produção de P3HB para se atingir uma produtividade de aproximadamente 1 g/L.h, considerando $\mu_{max} = 0,09 \text{ h}^{-1}$ e $q_{P3HB} = 32,3 \text{ mg/g.h}$.





Poly-3-hydroxybutyrate in
Recombinante *Escherichia coli*



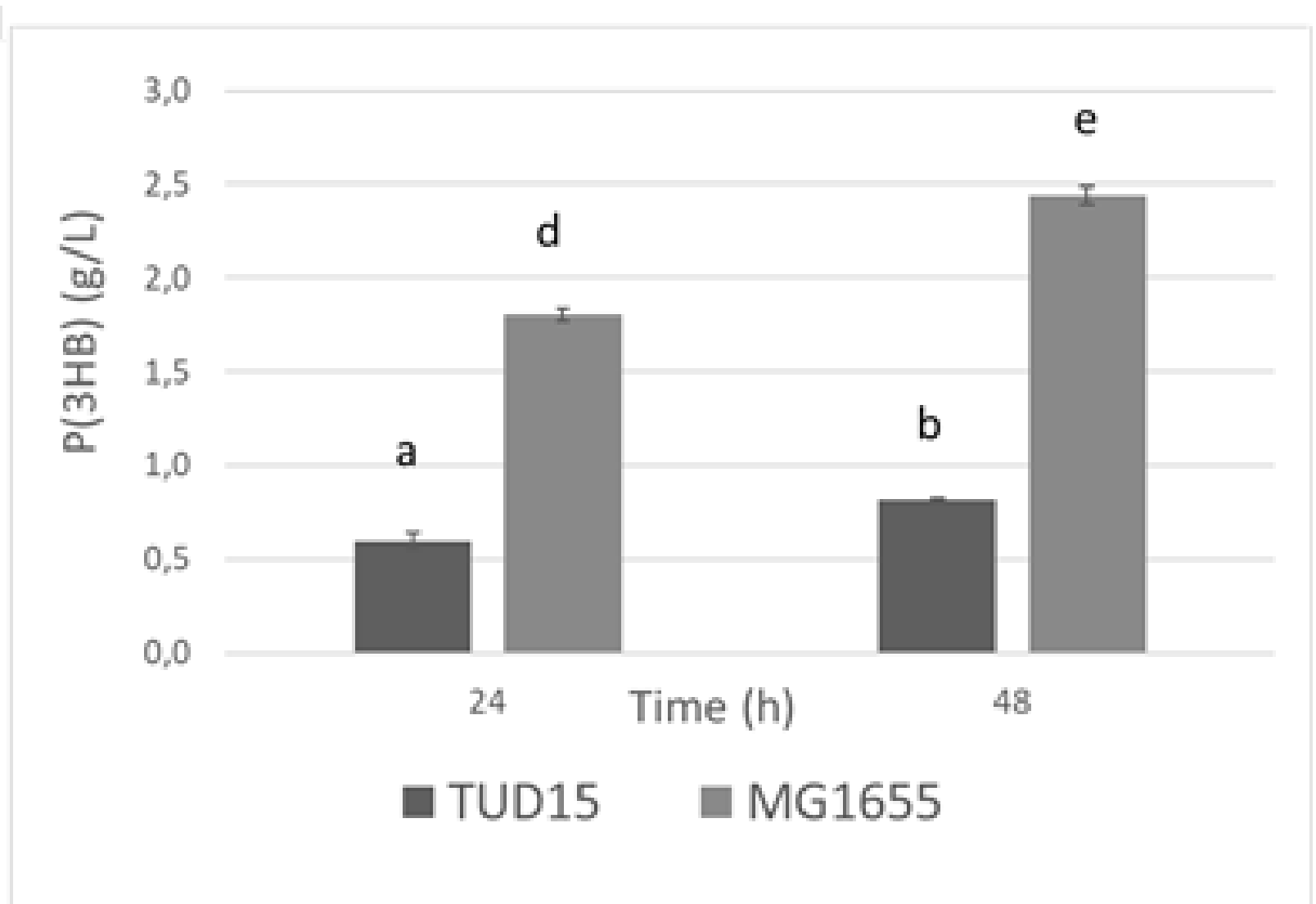


Figure 1. Production of P(3HB) by recombinant *E. coli* strains TUD15 and MG1655 (parental) harboring plasmid pBBR1MCS2:*phaCAB*. RCDW: residual cell dry biomass produced, CDW: total cell dry biomass produced.

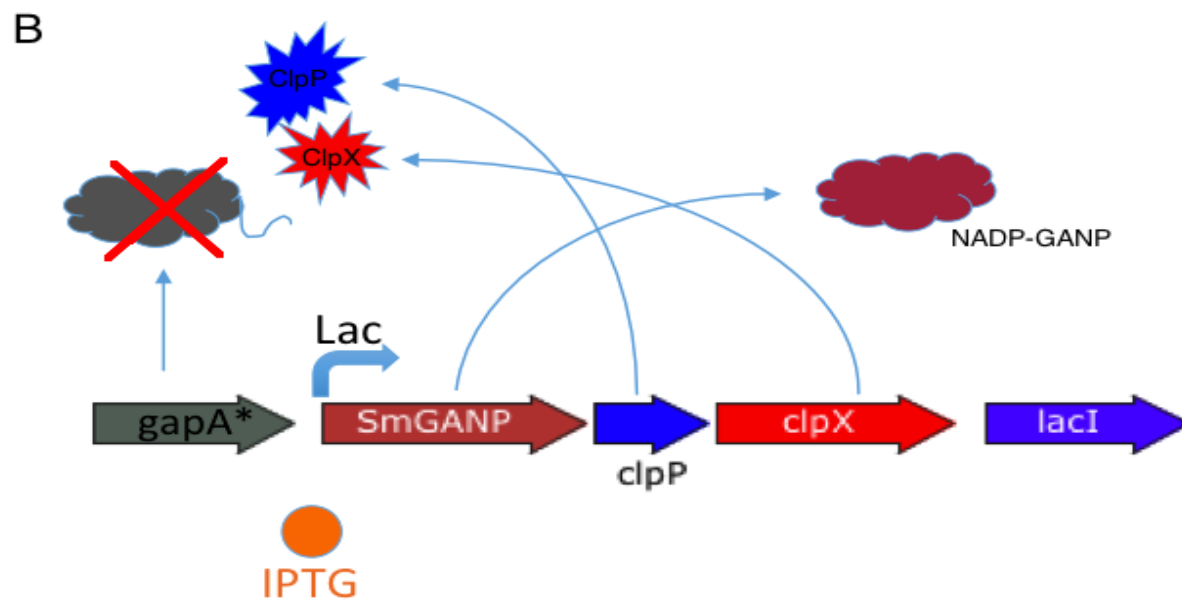
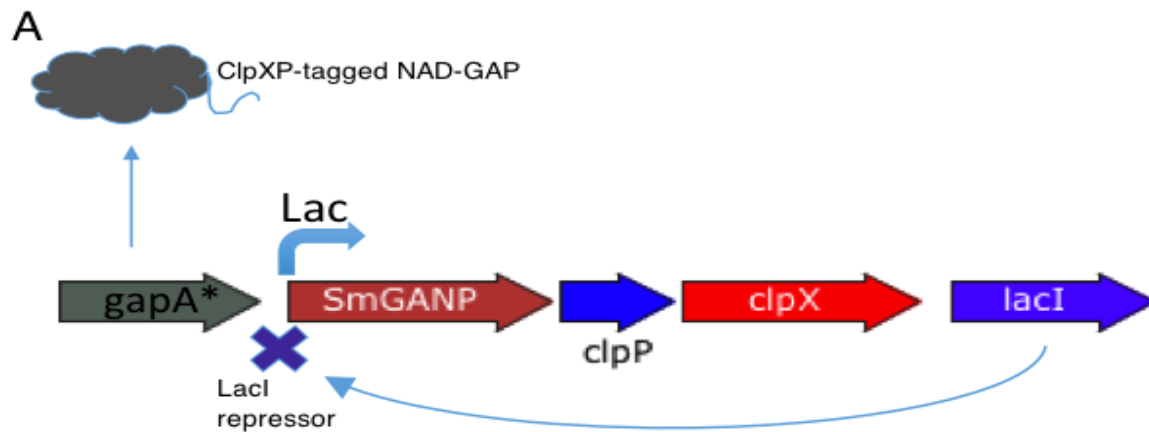


Table 2. Specific NAD-GAP and NADP-GANP activities measured in cellular extracts from the strains wild type (WT) and USP62. IPTG was added to the media at a final concentration of 1 mM. The cells were collected during the exponential growth phase. The values are average from three independent evaluations \pm standard deviations.

strains	^a NAD-GAP activity (mU/mg)	^b NADP-GANP activity (mU/mg)
WT no IPTG	1325 \pm 159	0
WT with IPTG	1346 \pm 147	0
USP62 no IPTG	219 \pm 21	0
USP62 with IPTG	727 \pm 21	36 \pm 8

^a 50 mM sodium phosphate (pH 7.0), 20 mM sodium arsenate, 4 mM glyceraldehyde-3-phosphate and 2 mM NAD.

^b 50 mM Tris (pH 8.5), 5 mM β -mercaptoethanol, 0.5 mM NADP, 1 mM glyceraldehyde-3-phosphate.

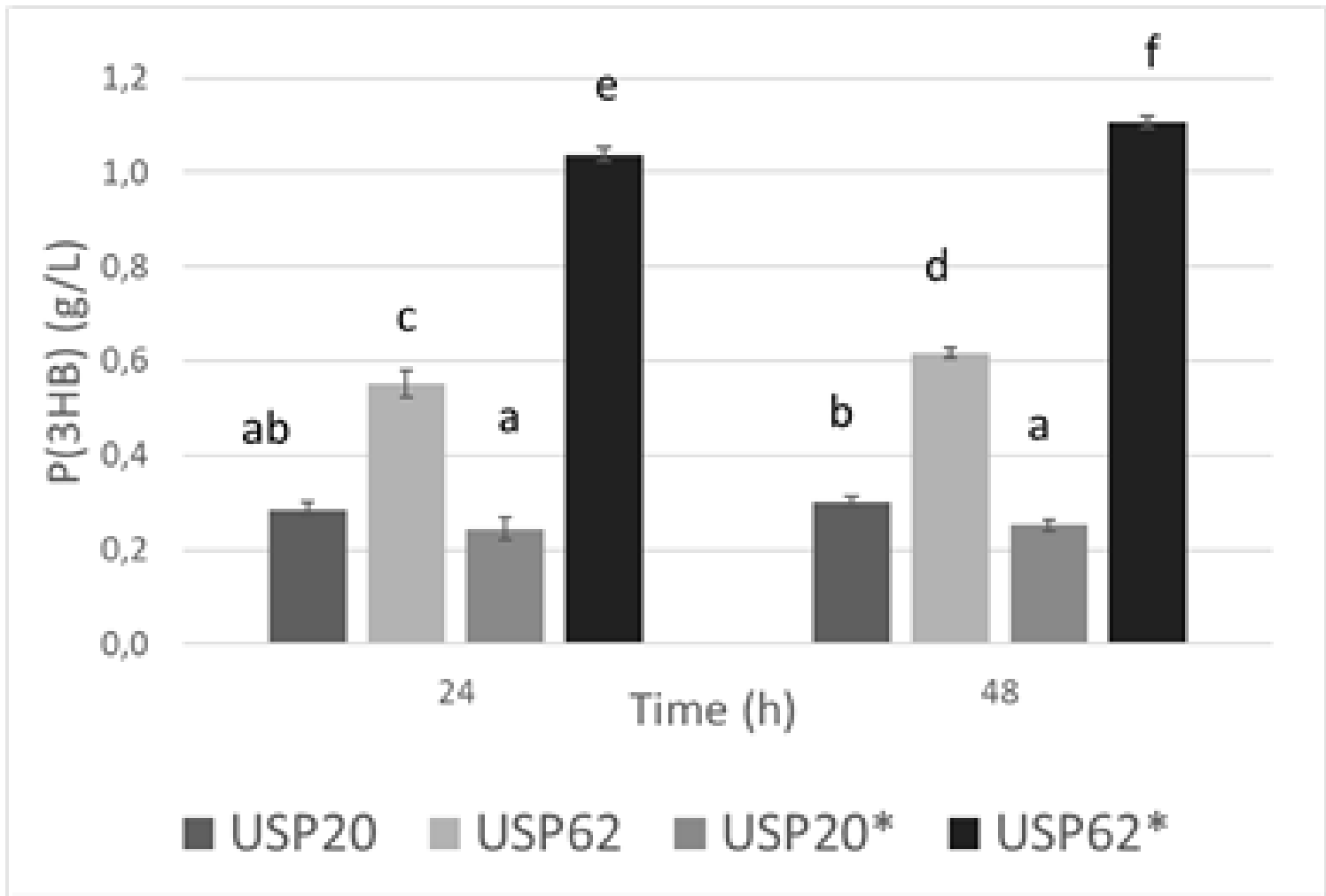
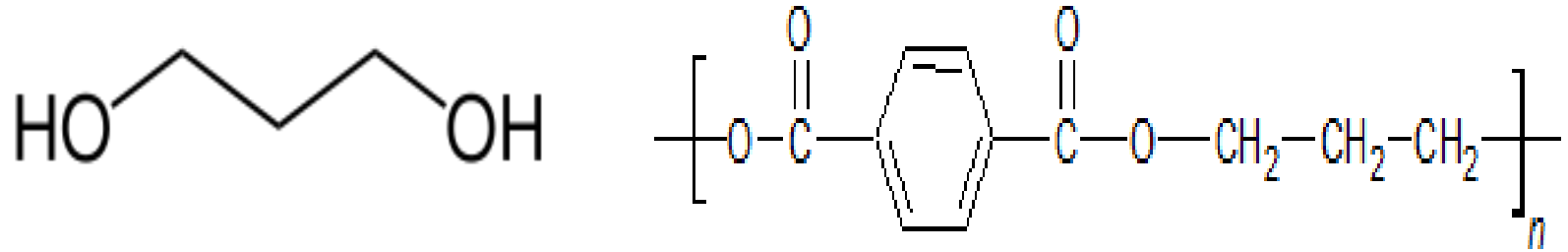


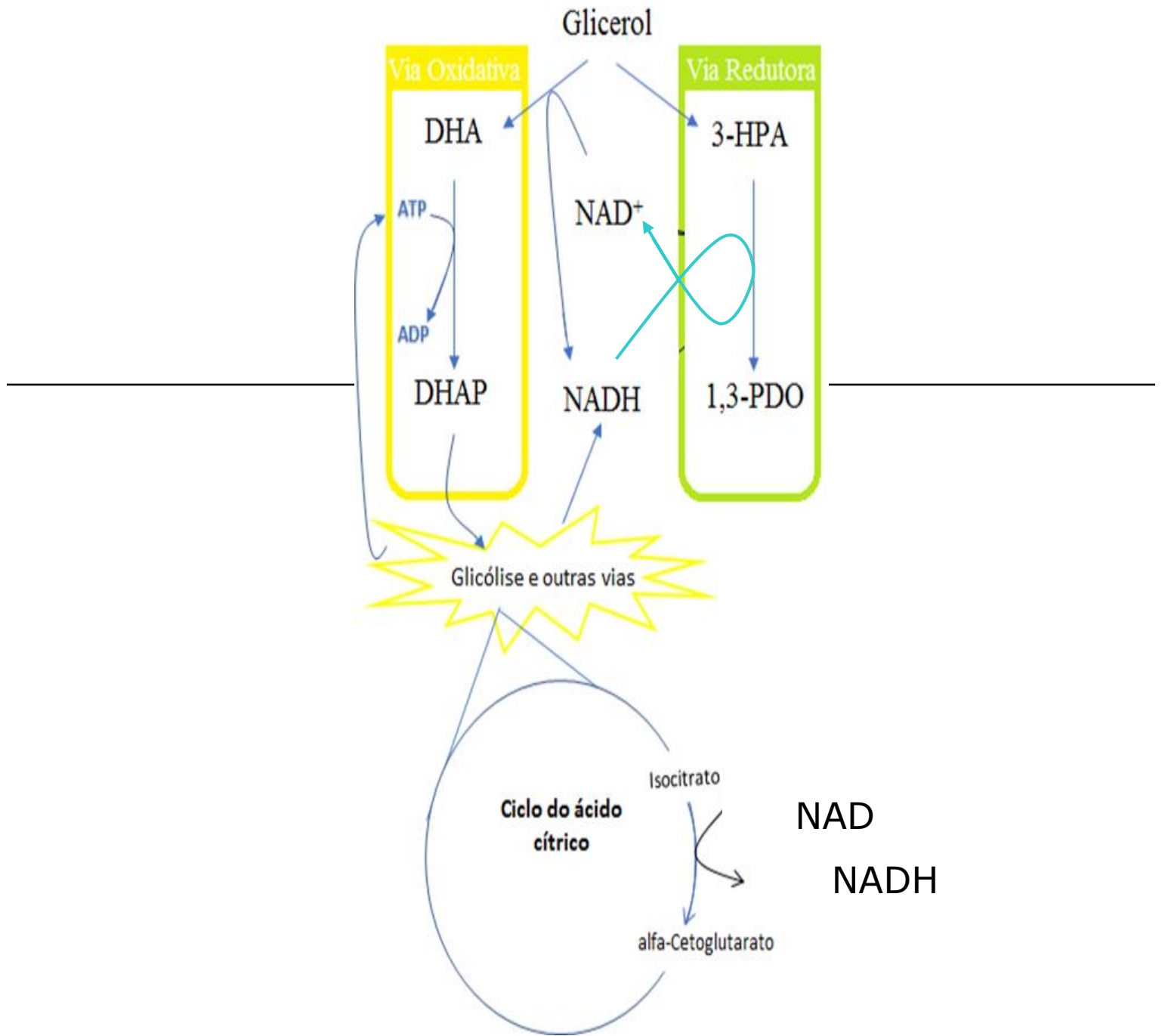
Figure 2. Production of P(3HB) by recombinant *E. coli* strains USP62 and USP20 (parental) harboring plasmid pBBR1MCS2:*phaCAB* with (*) and without the addition of IPTG. RCDW: residual cell dry biomass produced, CDW: total cell dry biomass produced.



1,3- propanediol production in Recombinante *Escherichia coli*

1,3-propanediol





Enzymatic activity of ICDH variants

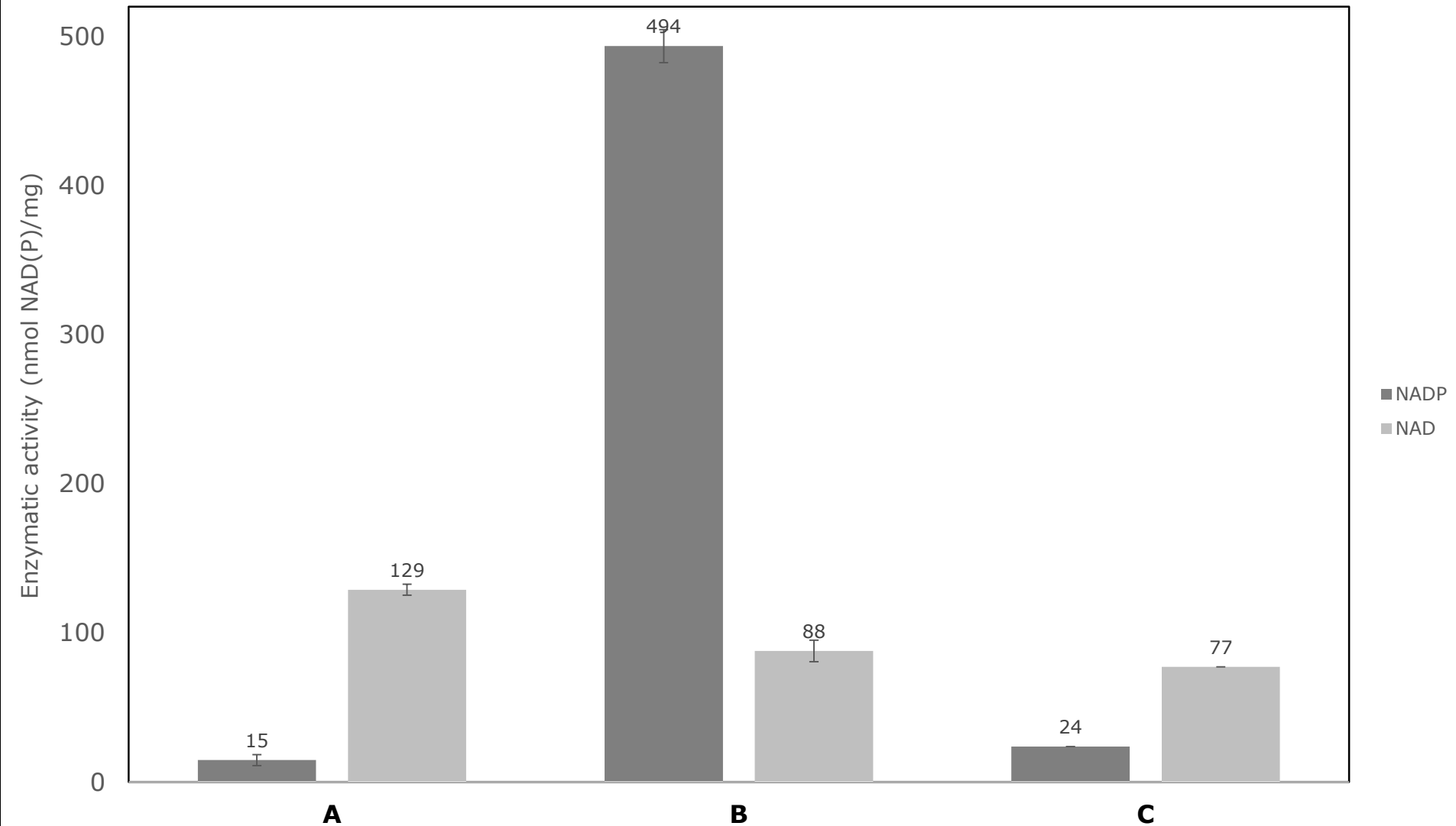
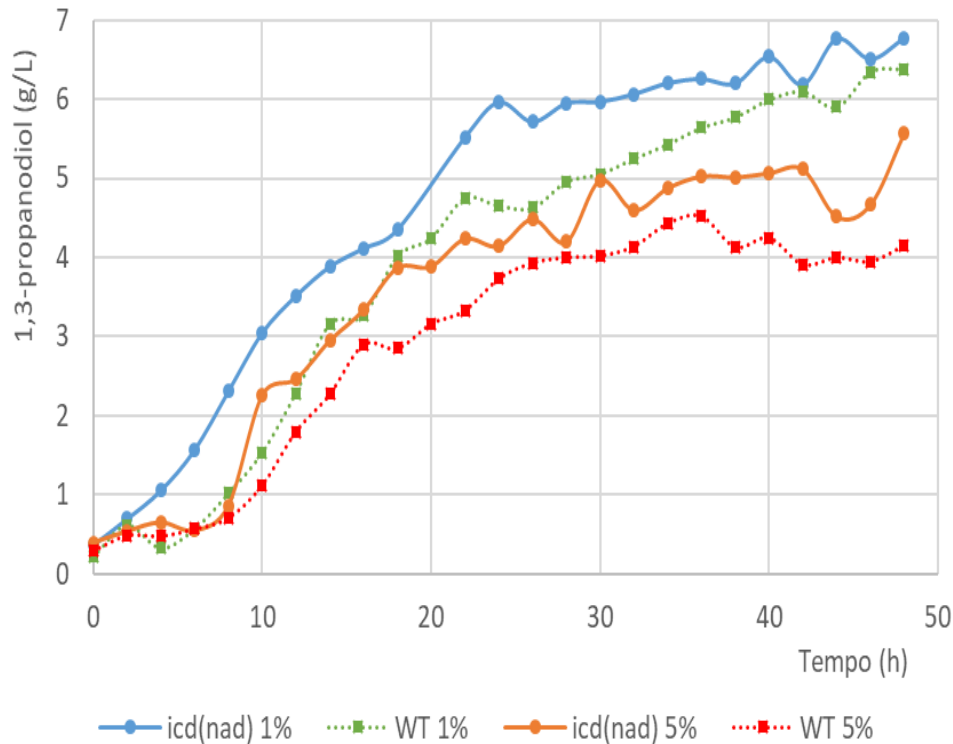


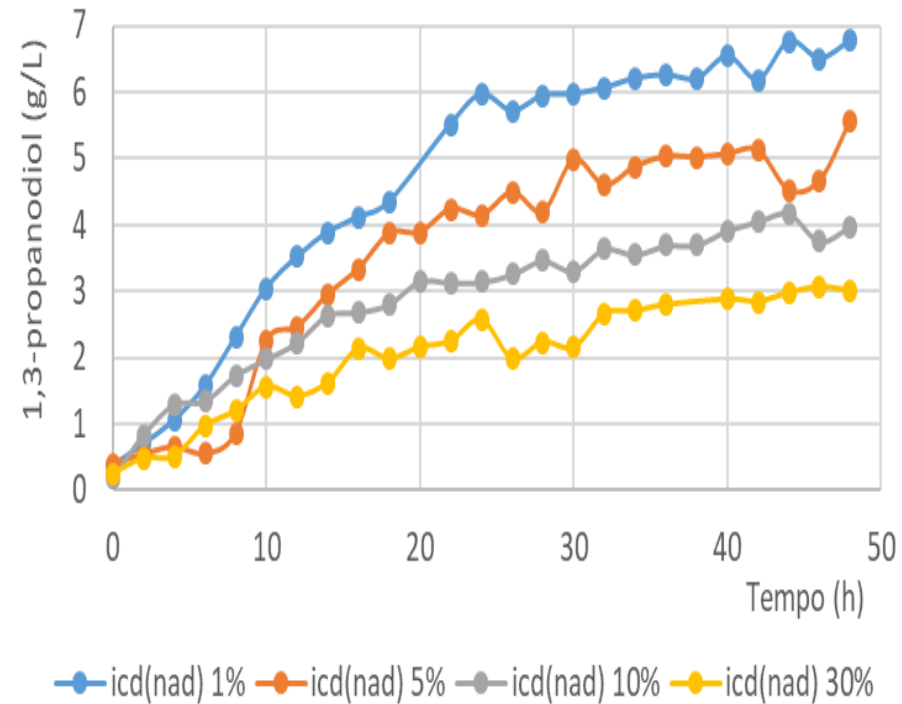
Figure 1:1 Enzymatic activity of ICDH variants. **A** = *E. coli* MG1655 Δicd (pTrc99A::*icd*^{NAD}) + 20 μ M de IPTG; **B** = *E. coli* MG1655; **C** = *E. coli* MG1655 *icd*^{NAD}. (■) = with NADP; (■) = with NAD.

○ Production of 1,3-PDO by Recombinant *E. coli*

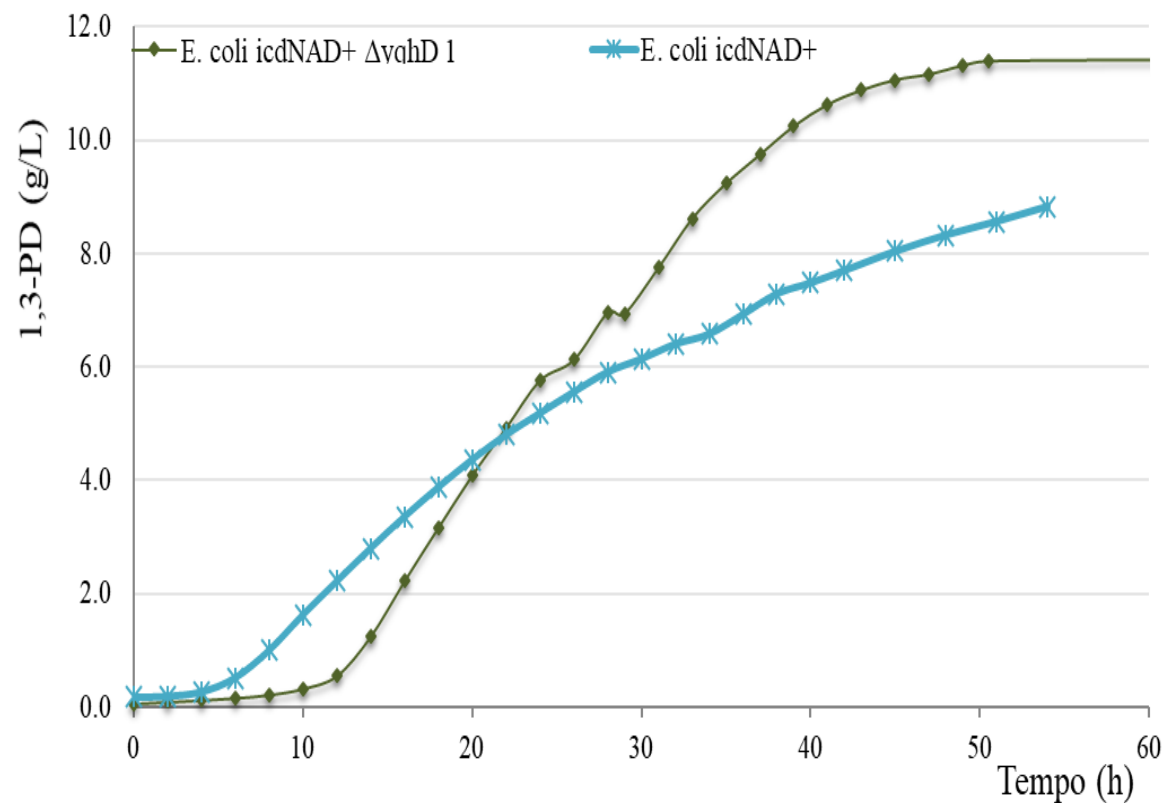
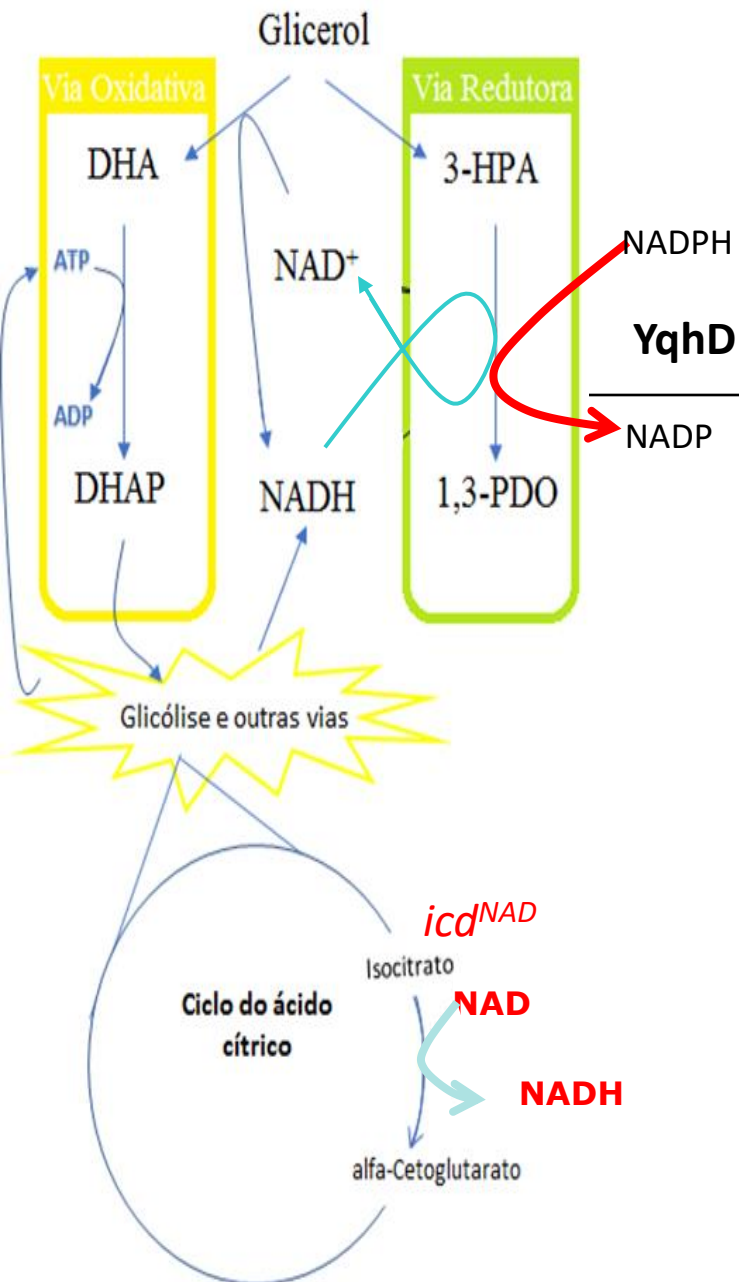
Produção de 1,3-PDO: icd WT e icd (nad)



Relação Oxigênio Dissolvido e produção de 1,3-PDO



Trabalhos anteriores



Global Transcription Machinery Engineering – Gtm Engineering

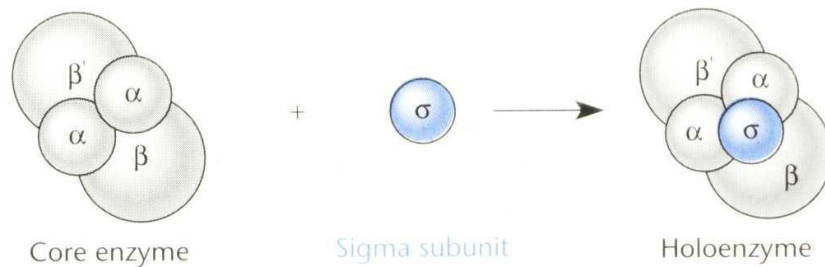


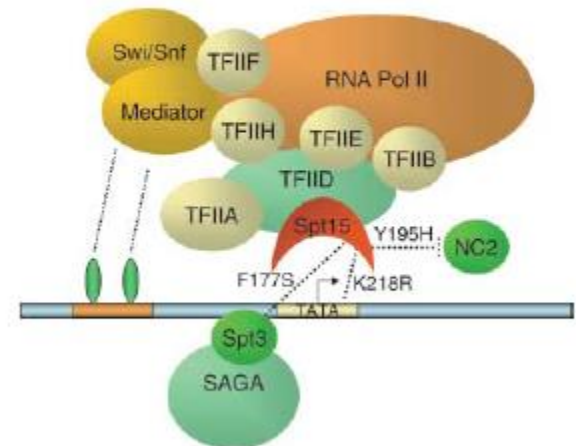
Figure 2.4 The composition of a typical bacterial RNA polymerase. The core enzyme contains two α subunits, a β subunit, and a β' subunit. The fifth subunit, the σ factor, cycles off after initiation of RNA synthesis.

Table 13.2 Some of the Alternative Sigma (σ) Factors in Bacteria and Their Cognate Promoter Sequences

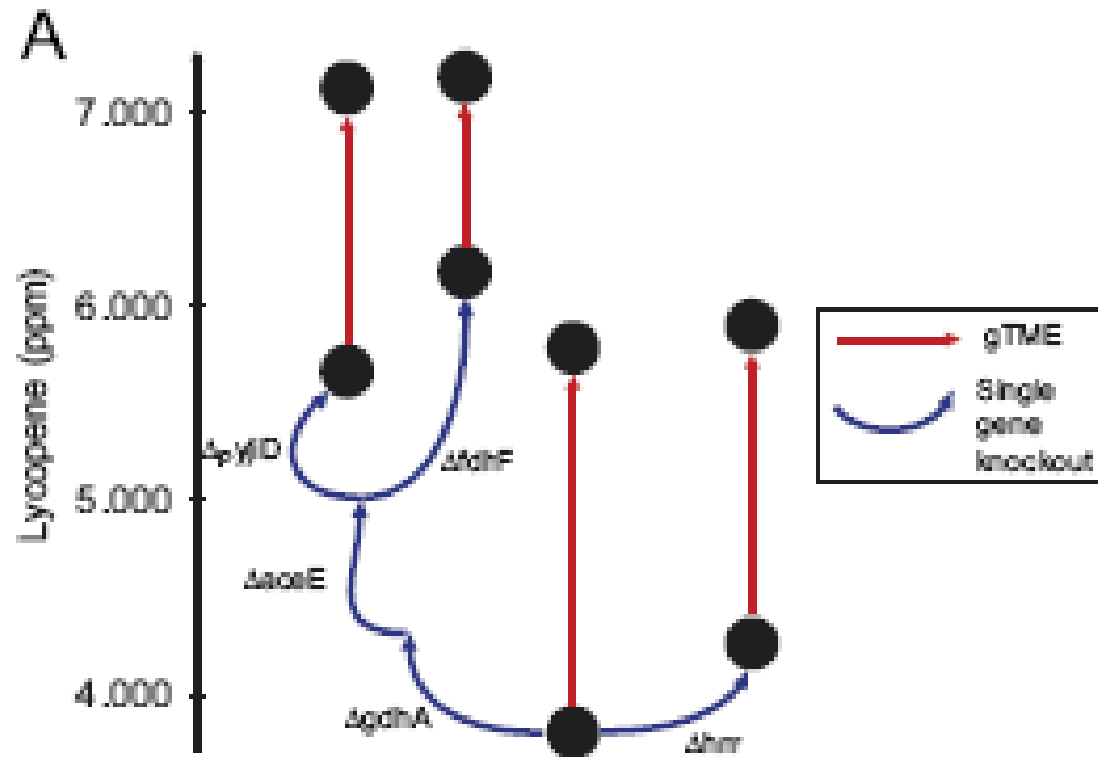
σ Factor Promoter Recognized	Genes Transcribed
σ^{70} TTGACA-17bp-TATAAT	Many and diverse
σ^{32} CNCITGAA-14bp-CCCCATNT	Heat shock response
σ^{54} CTGGNA-7bp-TTGCA	Many and diverse
σ^{28} TAA-15bp-GCCGATAA	Chemotaxis, motility, flagellar components
σ^{29} TTNAA-17bp-CATATT	Sporulation in <i>B. subtilis</i> ¹

¹Several different sigma factors control different genes involved in sporulation in *B. subtilis*.

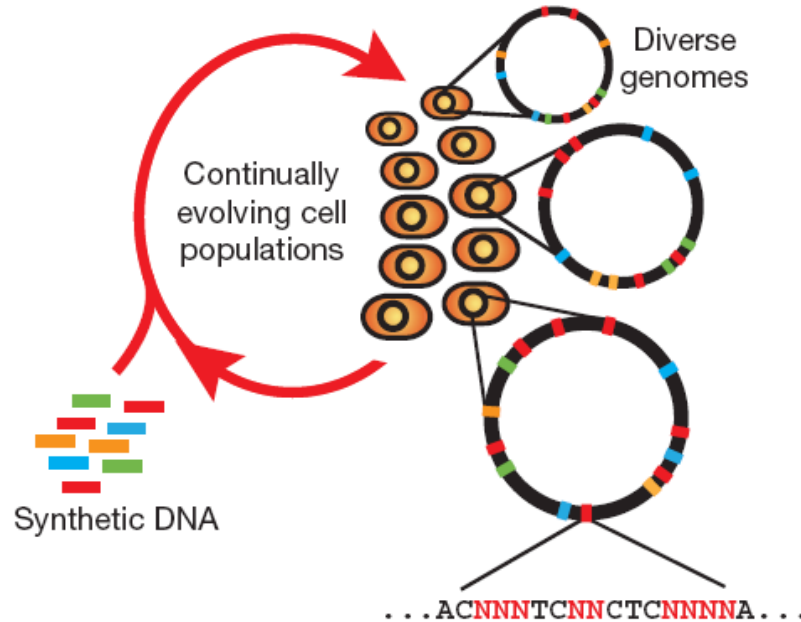
B



Global Transcription Machinery Engineering – gTME



MAGE



With the advent of next-generation fluorescent DNA sequencing¹⁰, our ability to sequence genomes has greatly outpaced our ability to modify genomes. Existing cloning-based technologies are confined to serial and inefficient introduction of single DNA constructs into cells, requiring laborious and outdated genetic engineering techniques. Whereas *in vivo* methods such as recombination-based genetic engineering (recombineering) have enabled efficient modification of single genetic targets using single-stranded DNA (ssDNA)^{11–14}, no such attempts have been made to modify genomes on a large and parallel scale. MAGE provides a highly efficient, inexpensive and automated solution to simultaneously modify many genomic locations (for example, genes, regulatory regions) across different length scales, from the nucleotide to the genome level (Fig. 1).

Figure 1 | Multiplex automated genome engineering enables the rapid and continuous generation of sequence diversity at many targeted chromosomal locations across a large population of cells through the repeated introduction of synthetic DNA. Each cell contains a different set of mutations, producing a heterogeneous population of rich diversity (denoted by distinct chromosomes in different cells). Degenerate oligo pools that target specific genomic positions enable the generation of a diverse set of sequences at each chromosomal location.

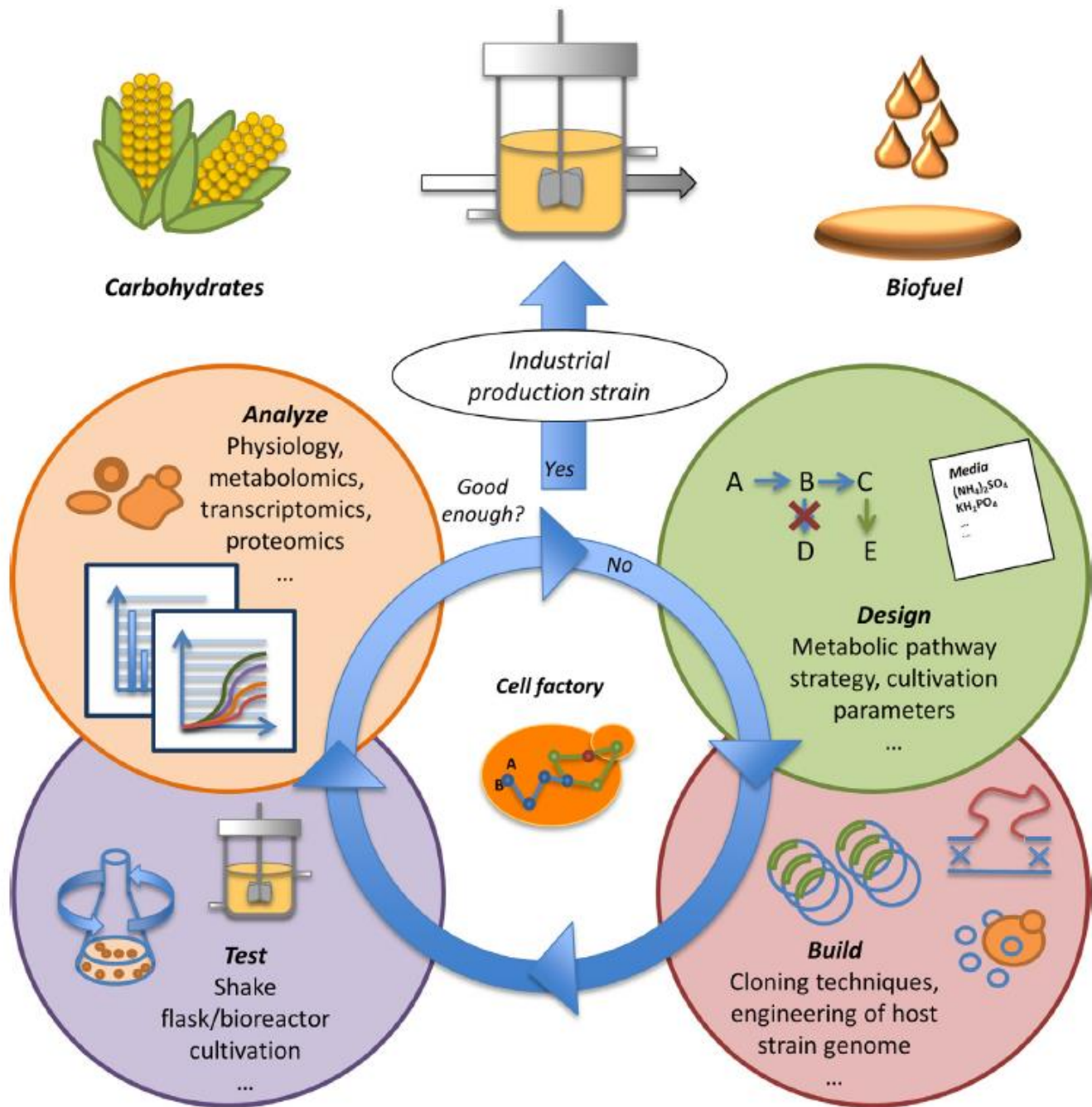


FIGURE 7.1 The iterative metabolic engineering cycle used to construct cell factories.

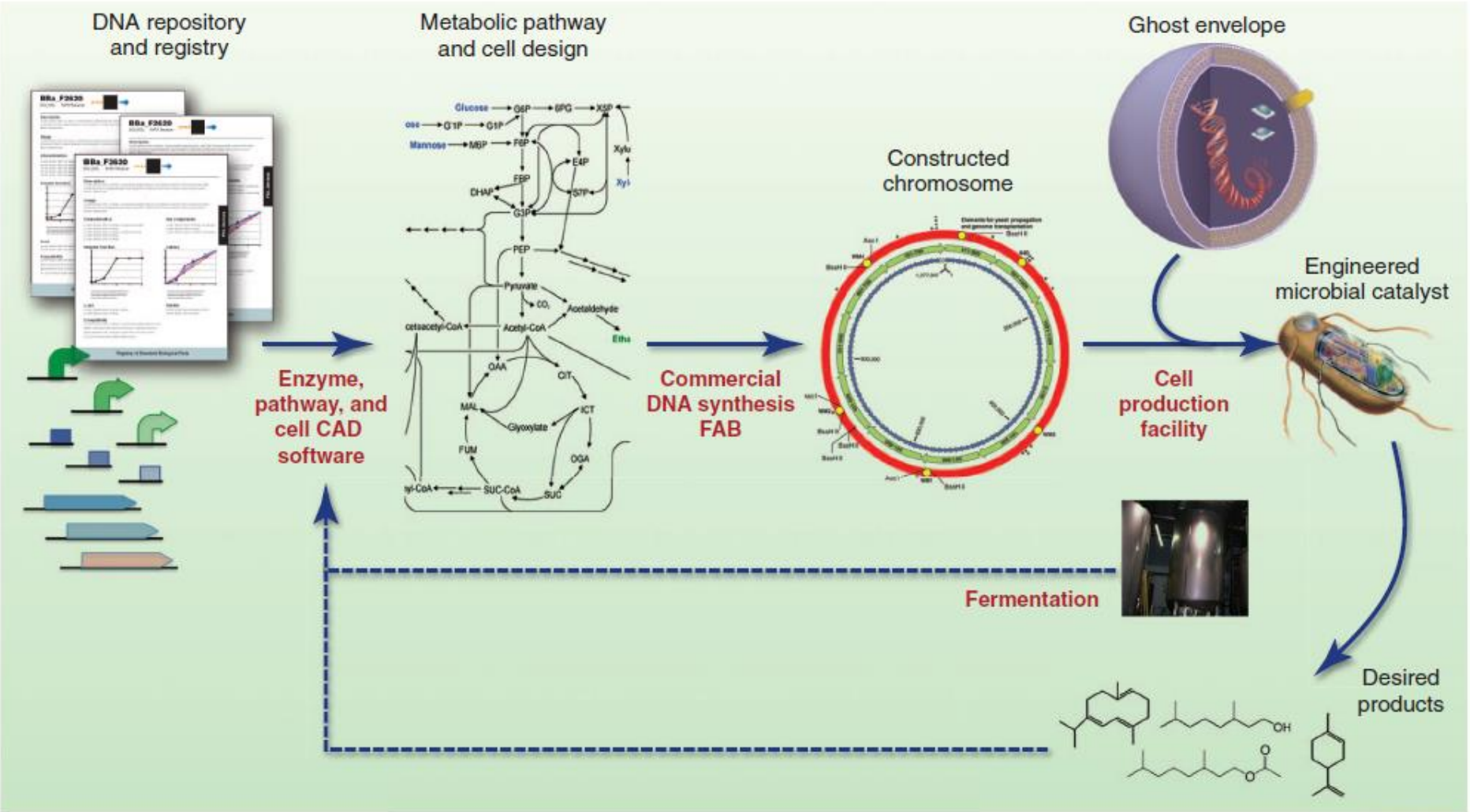


Fig. 3. The future of engineered biocatalysts. Pathways, enzymes, and genetic controls are designed from characteristics of parts (enzymes, promoters, etc.) by means of pathway and enzyme CAD software. The chromosomes encoding

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.

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