

# Co-inoculation of bacterial strains and data analysis

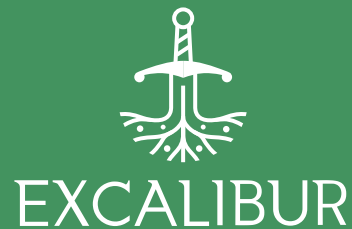
April 19<sup>th</sup>, 2023

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EXCALIBUR Training Series:

Addressing microbial metabolic profile by means of Phenotype Microarray technology (BIOLOG)



This project has received funding from the European Union's Horizon 2020 research and innovation programme under grant agreement No 817946

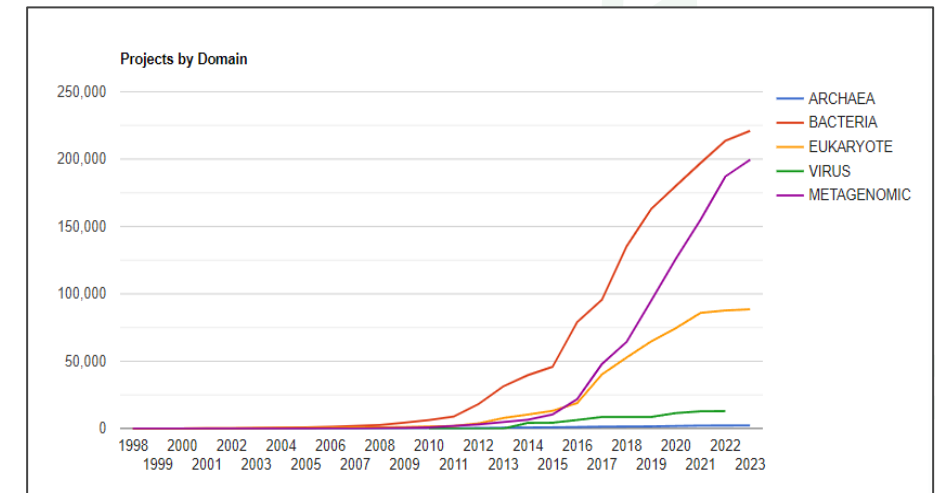
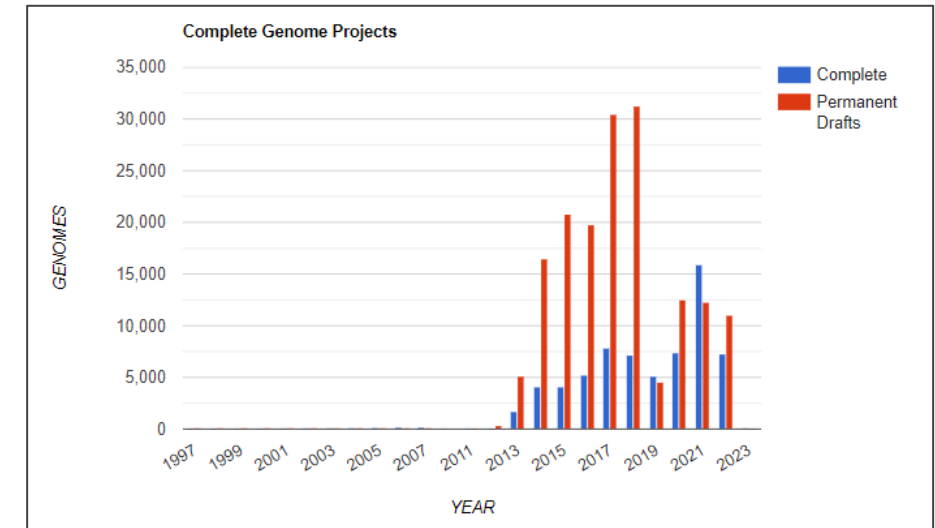


# The genomic era

Addressing the functionality of genomes is one of the most important and challenging tasks of today's biology. This, together with the impressive improvements of sequencing technology and bioinformatic tools, have provided a huge enhancement of the number of genomic projects over the world.

However, even though such techniques are of great help, **the final expression of the genomic information is represented by the phenotype**

- Metabolic networks reconstruction
- From genomes to metabolomes
- High throughput genomics/metabolomics





# The phenomic era

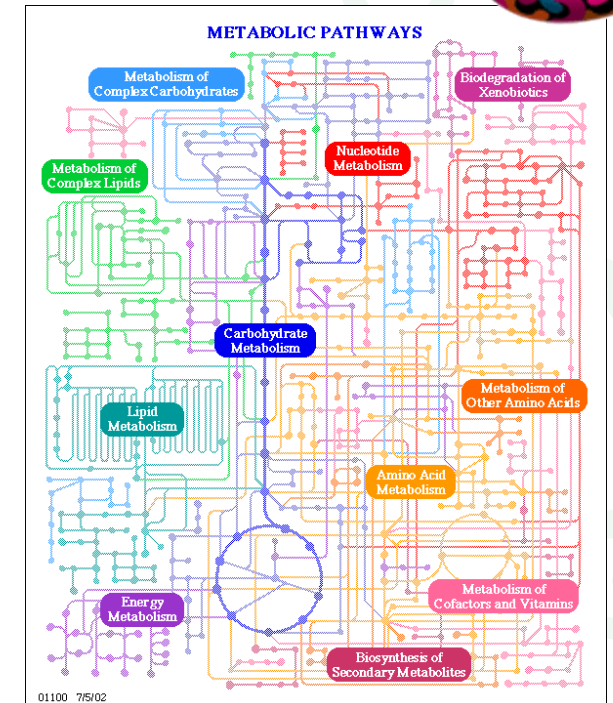
The OmniLog™ platform, uses 96-well PM plates, composed of more than 2000 temperature-controlled conditions.

In each well the metabolism of the cell can be monitored using respiration as a reporter system

- Up to 20 plates (1920 wells)
- Up to 9 compound categories
- Time series data
- Many compounds on KEGG DB
- High throughput phenomics

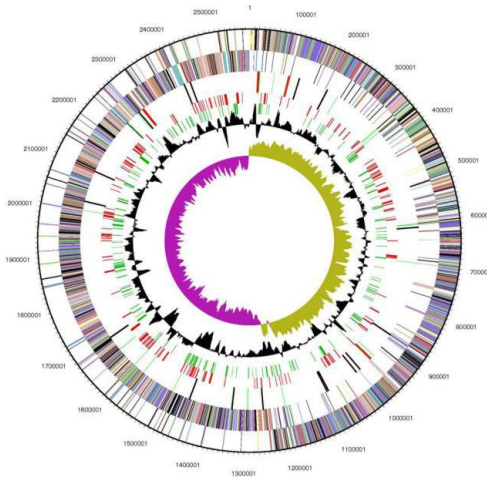


[www.biolog.com](http://www.biolog.com)



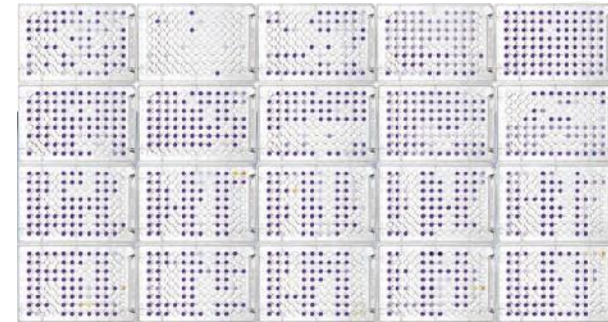
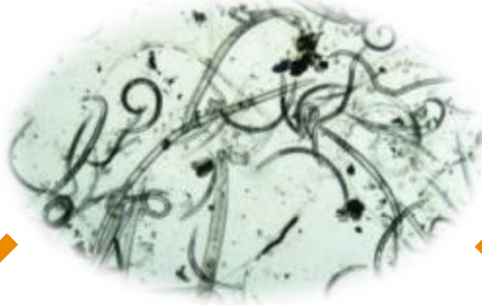
# The 'omic' era

Genome and phenome data can be achieved from soil bacteria



## Genome data analysis

- Genome map to KEGG
- Pangenome prediction
- core
- accessory
- unique

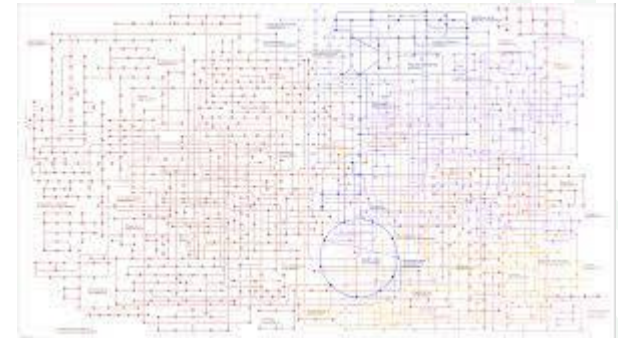
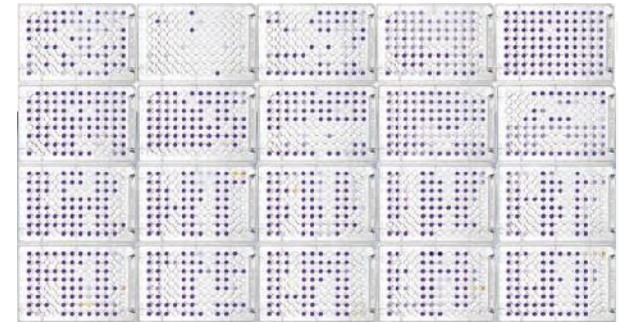
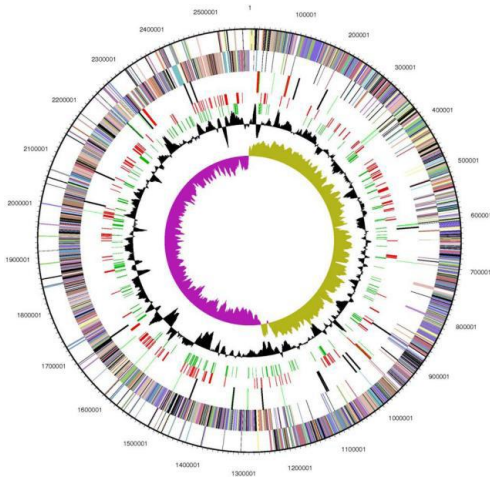


## Phenome data analysis

- Metabolic activity parameters
- Replica management
- Clear comparisons
- Clear visualizations
- Compounds map to KEGG

# The 'omic' era

**But...which genes may be responsible for such phenotypic variations?**

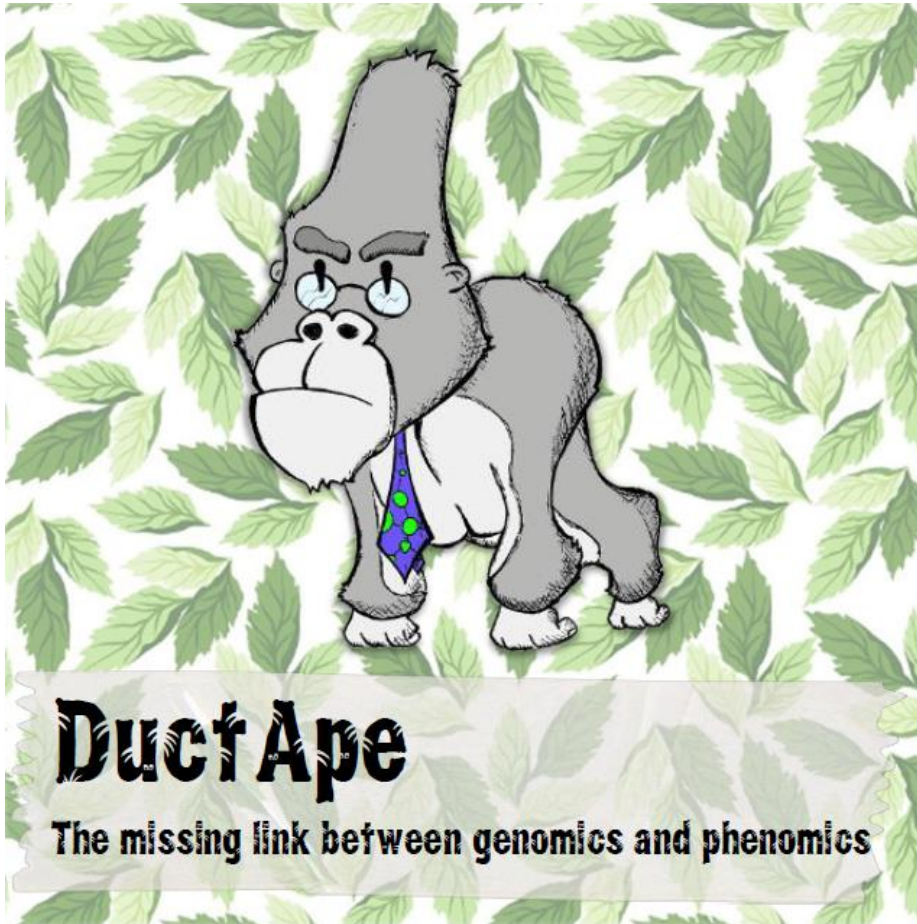


**How to combine genomic and phenomic data?**

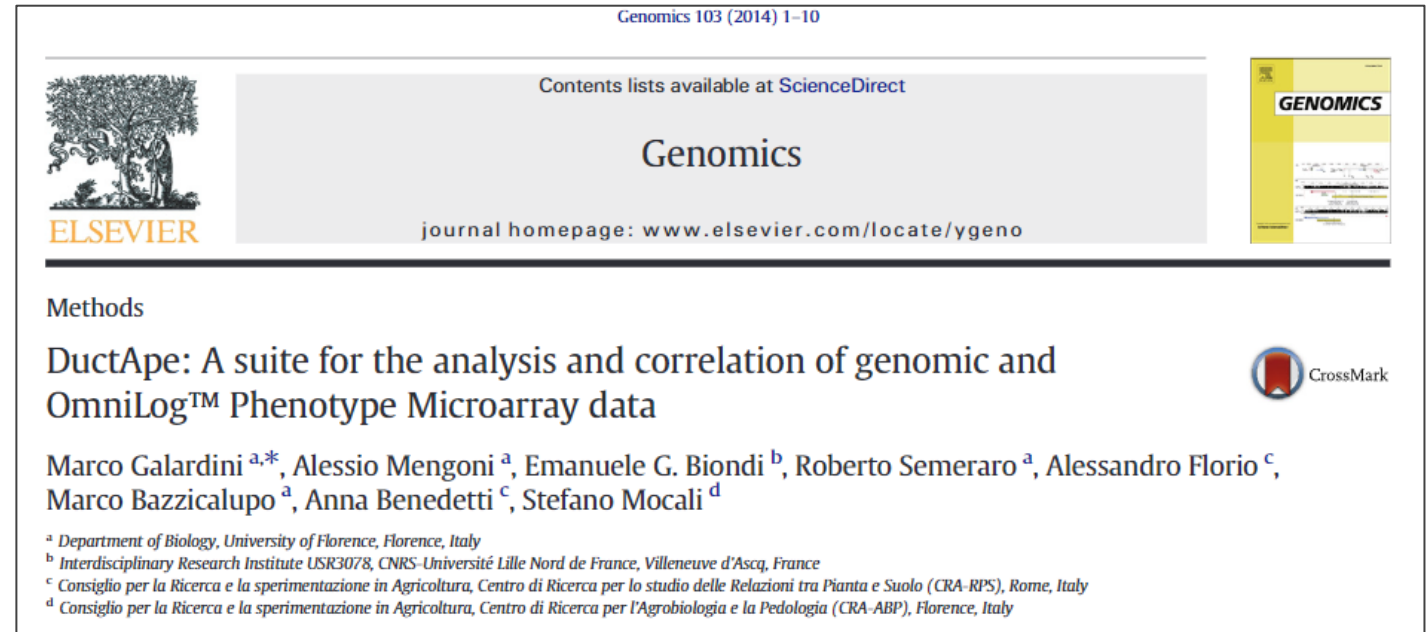
- All data in a single metabolic map
- Genetic basis for phenotypic differences



# Ductape software

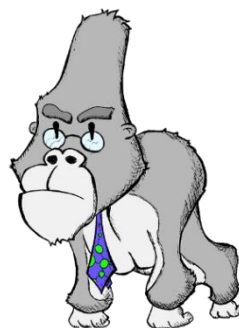
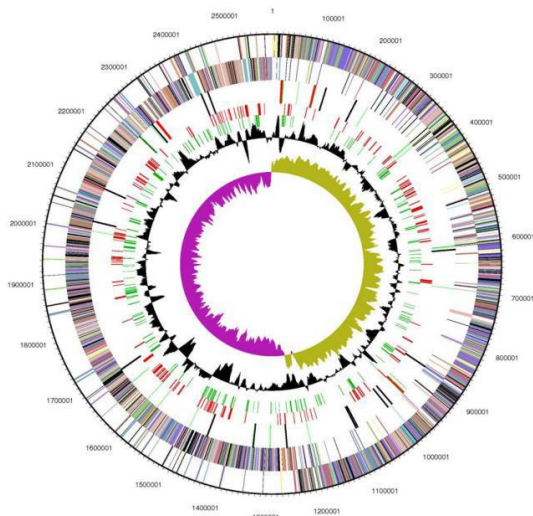


<http://combogenomics.github.com/DuctApe>



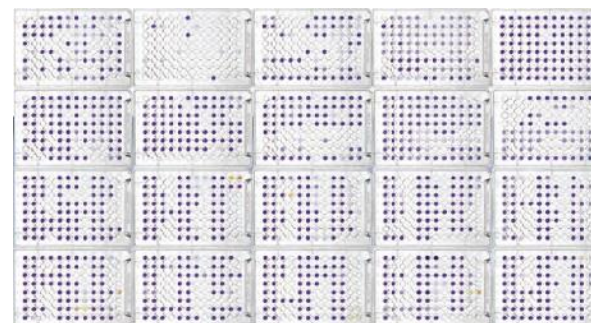
- Reconstruct metabolic maps from proteome sequences (derived from genomes)
- Interpret Phenotype Microarray (BIOLOG) curves
- Provide testable hypothesis for genotype-to-phenotype links
- Available as a command line tool (not in active development) for Unix systems (Linux/macOS) (on Windows you can use a VM)

# Ductape software



## dape

- Generation of combined KEGG metabolic maps
- Metabolic network analysis
- Metabolic hotspots prediction



## dgenome

- Genes are mapped to KEGG database
- PanGenome prediction (Blast-BBH)

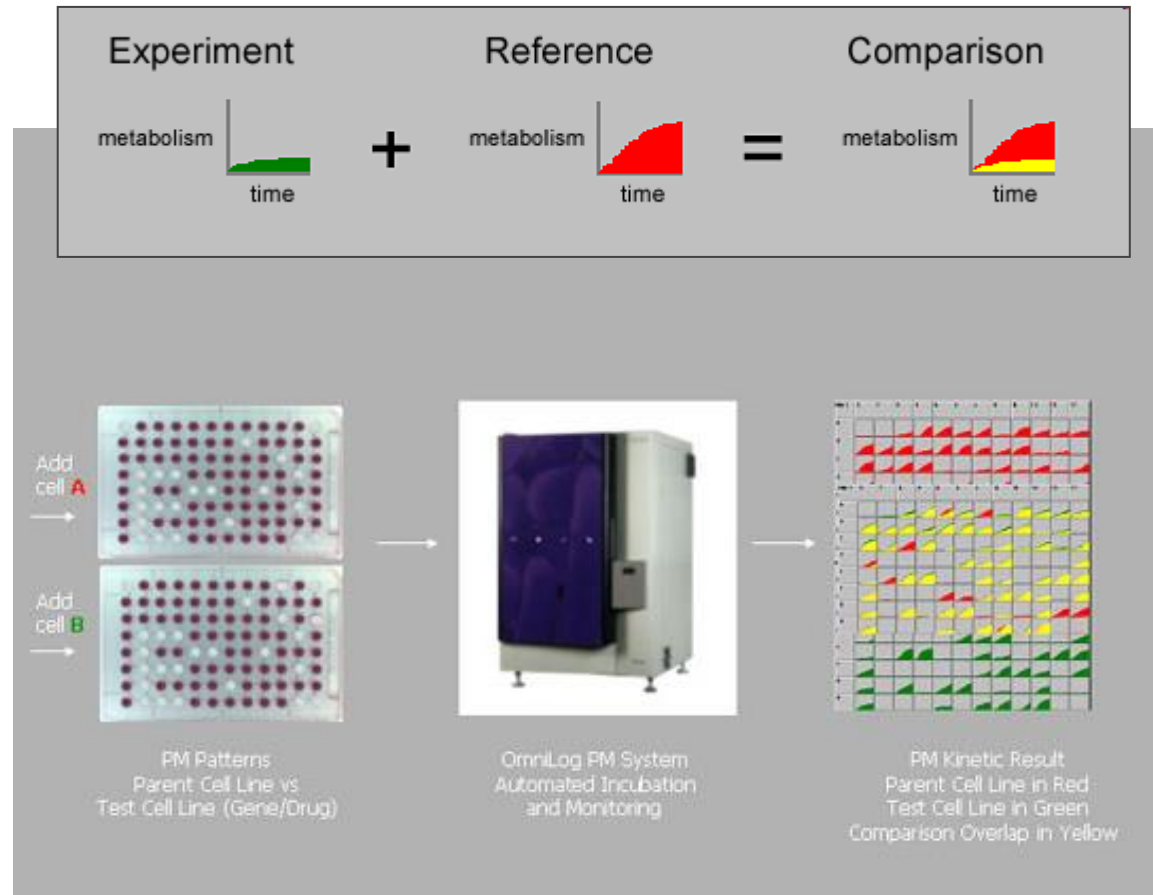
## dphenome

- Phenotype microarray data handling
- Classification of metabolic activity
- Compounds are mapped to KEGG DB

# Comparison of phenotypes

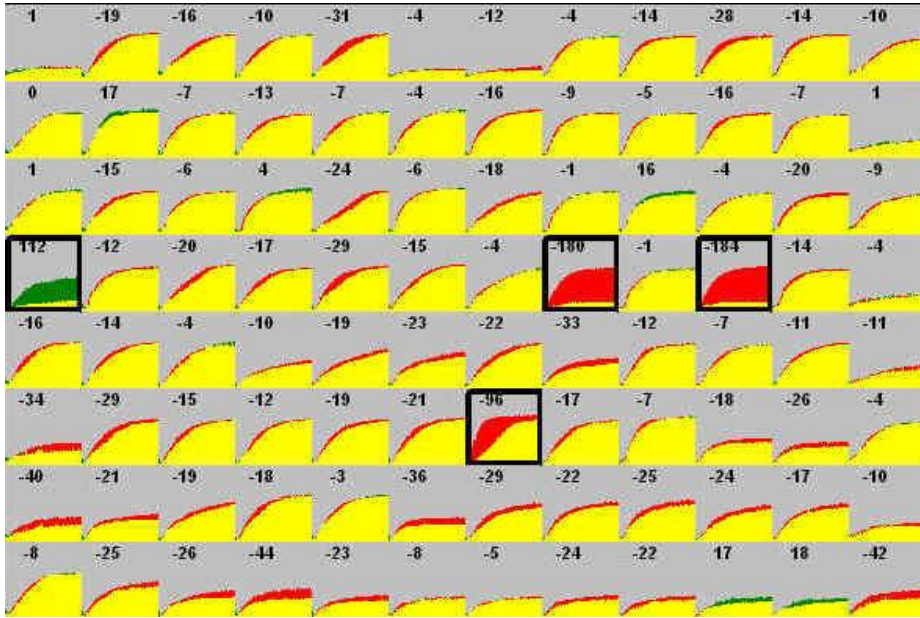
To compare the phenotypes of two cell lines or strains, one is recorded as a **RED** tracing and one as a **GREEN** tracing. These graphs can then be overlaid by the bioinformatic software to detect differences.

Areas of overlap (i.e. no change) are colored **YELLOW**, whereas differences are highlighted as patches of red or green

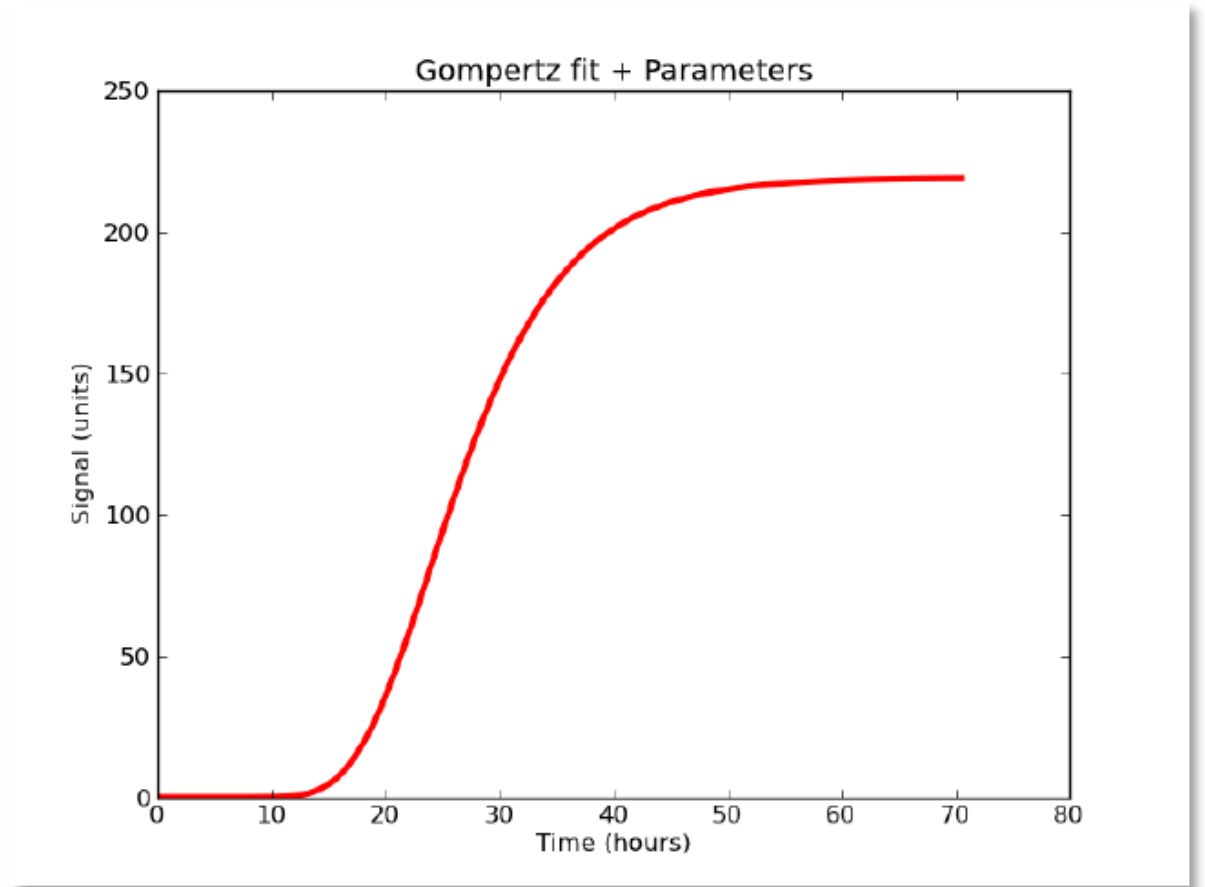


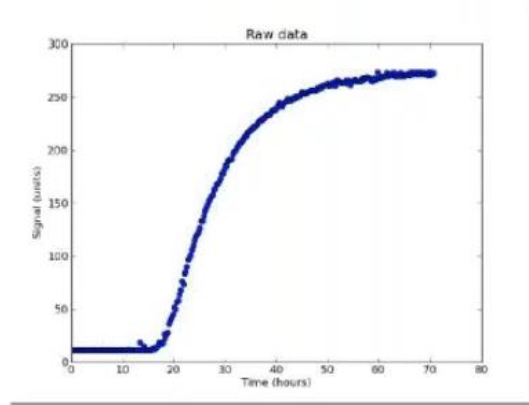


# Comparison of phenotypes

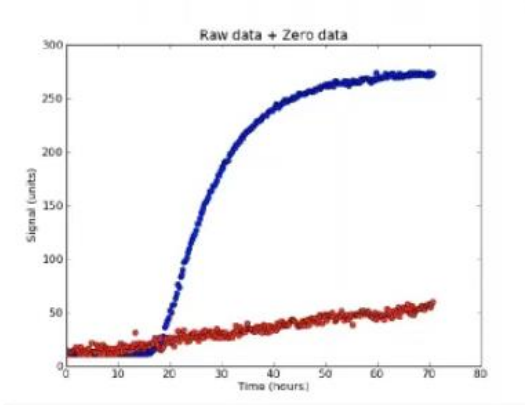


- Multiple strain comparison
- How to discriminate different activities?
- A single, summarized value is needed

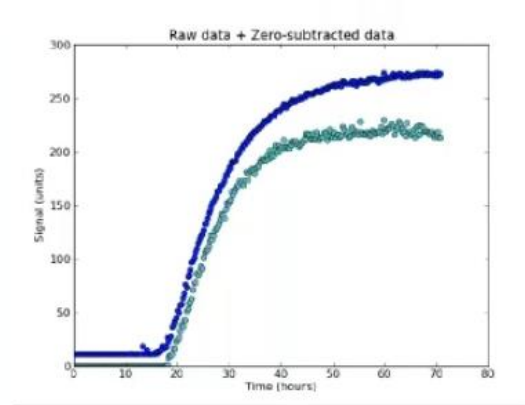




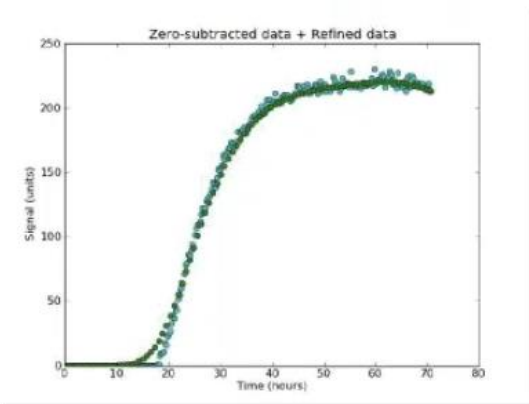
1. Parsing



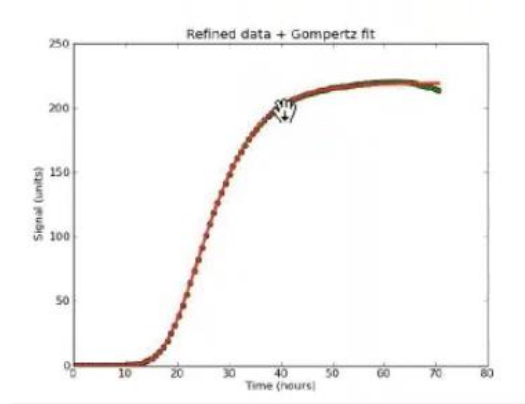
2. Control signal subtraction (optional)



3. Signal refinement



4. Sigmoid fit



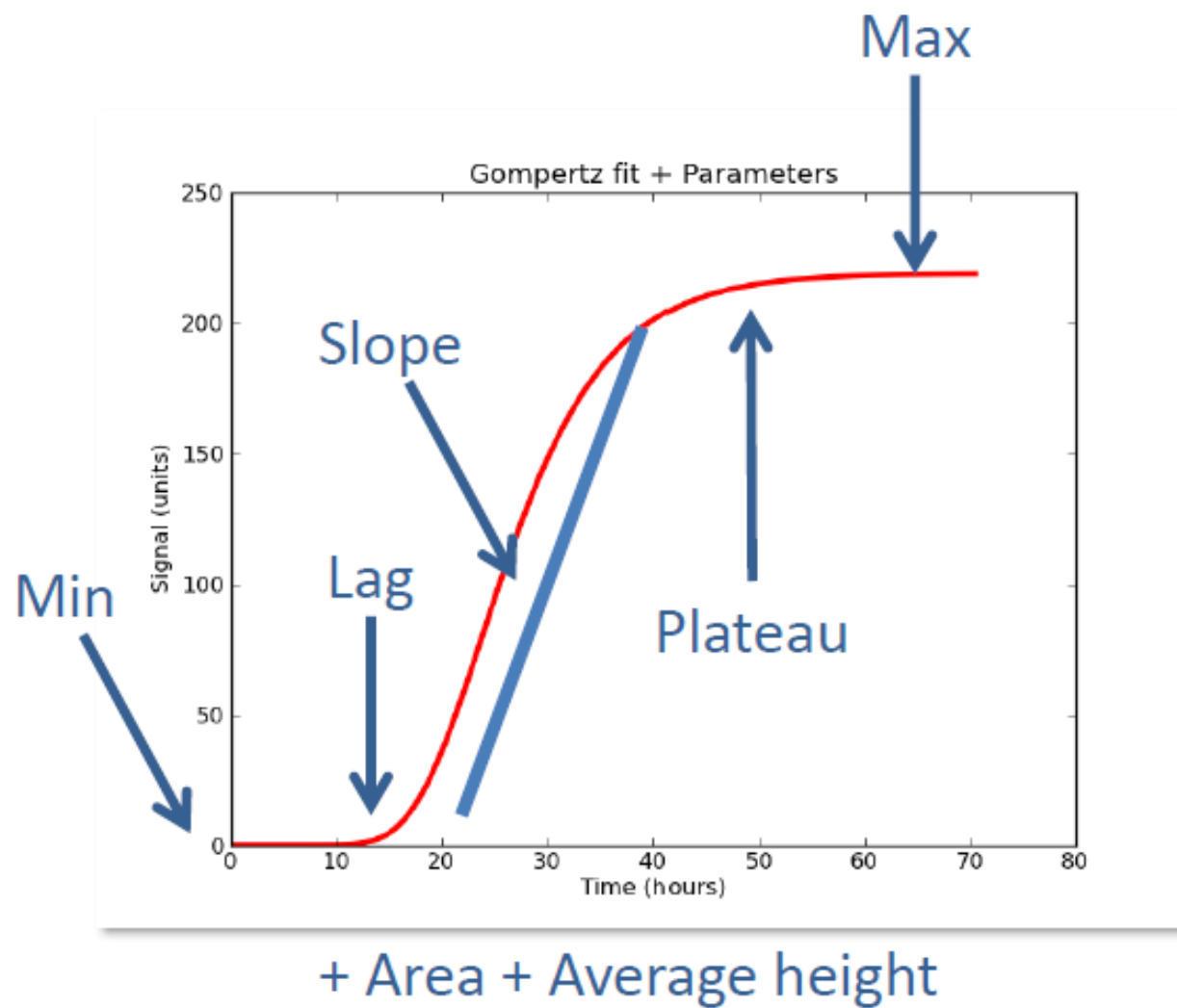
## Modeling of the Bacterial Growth Curve

M. H. ZWIETERING,\* I. JONGENBURGER, F. M. ROMBOUTS, AND K. VAN 'T RIET  
Department of Food Science, Agricultural University Wageningen, P.O. Box 8129,  
6700 EV Wageningen, The Netherlands

Received 5 January 1990/Accepted 4 April 1990

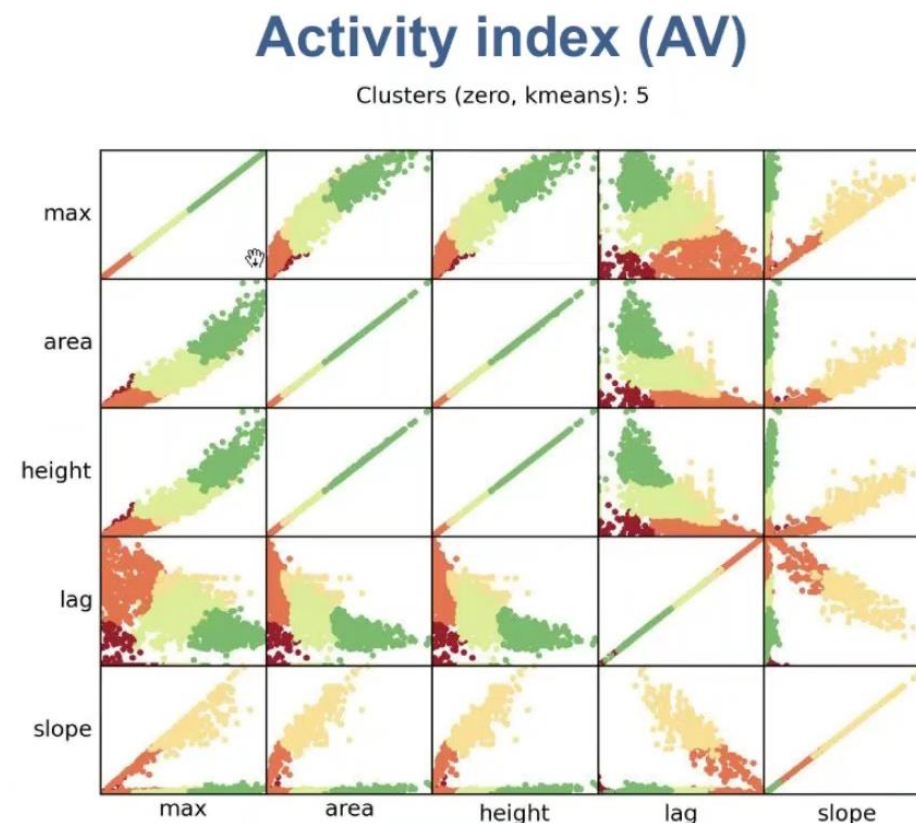
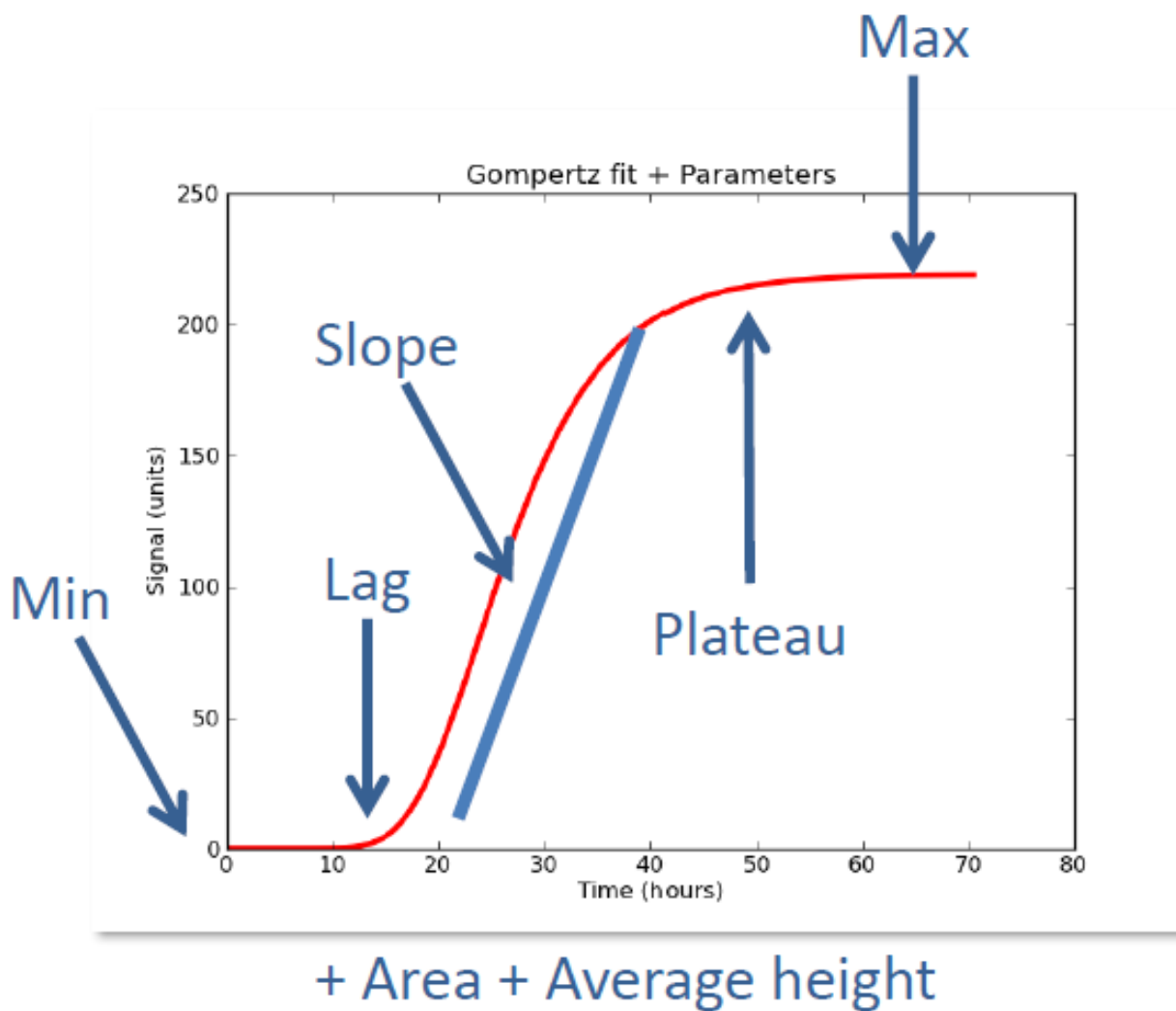
Model	Equation
Logistic	$y = \frac{a}{[1 + \exp(b - cx)]}$
Gompertz	$y = a \cdot \exp[-\exp(b - cx)]$
Richards	$y = a \{1 + v \cdot \exp[k(\tau - x)]\}^{(-1/v)}$

# Activity index (AV)





# Activity index (AV)

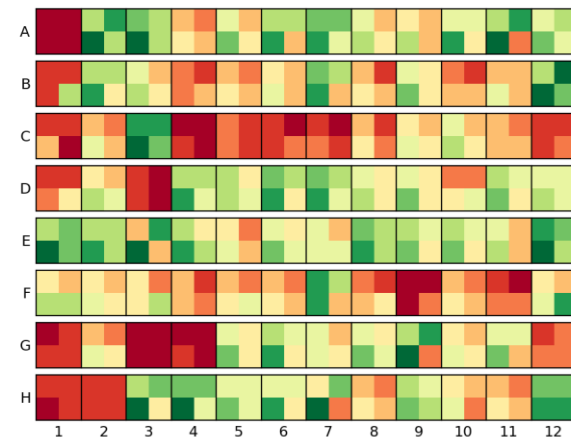
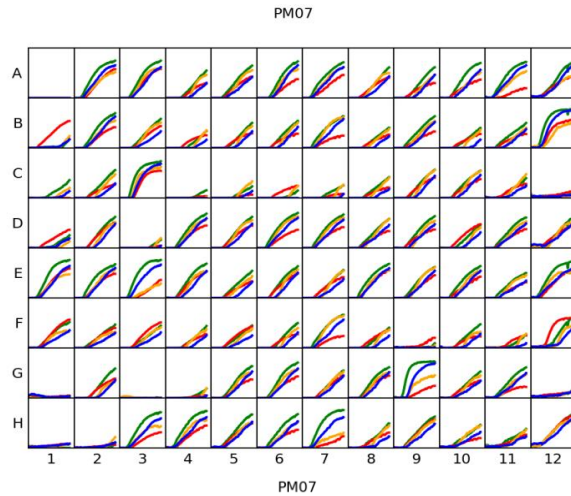


K-means clustering on 5 parameters  
(5 clusters in this example, tunable by the user)

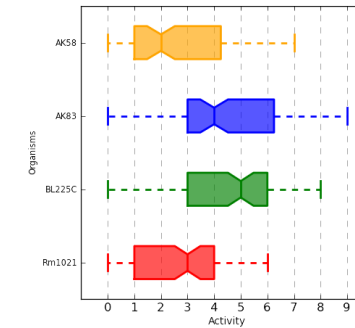
**Fast:** from raw .csv files to AV in less than 5 minutes

# From raw data to phenotypic variability

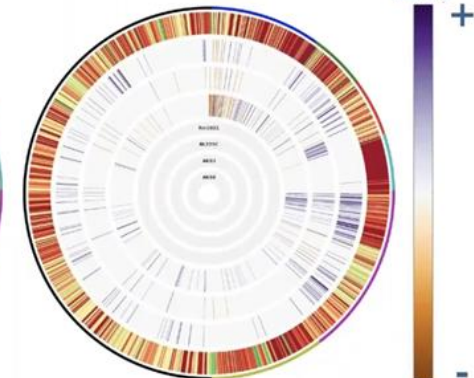
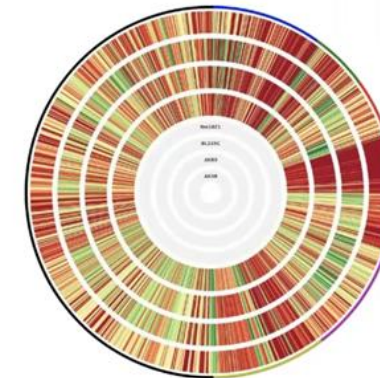
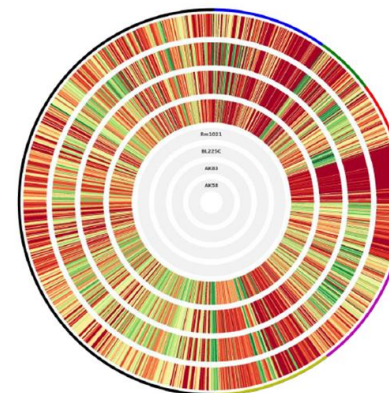
**Plates AV heatmaps:** phenotypic variability at a glance



**AV boxplots:** overall strains comparison

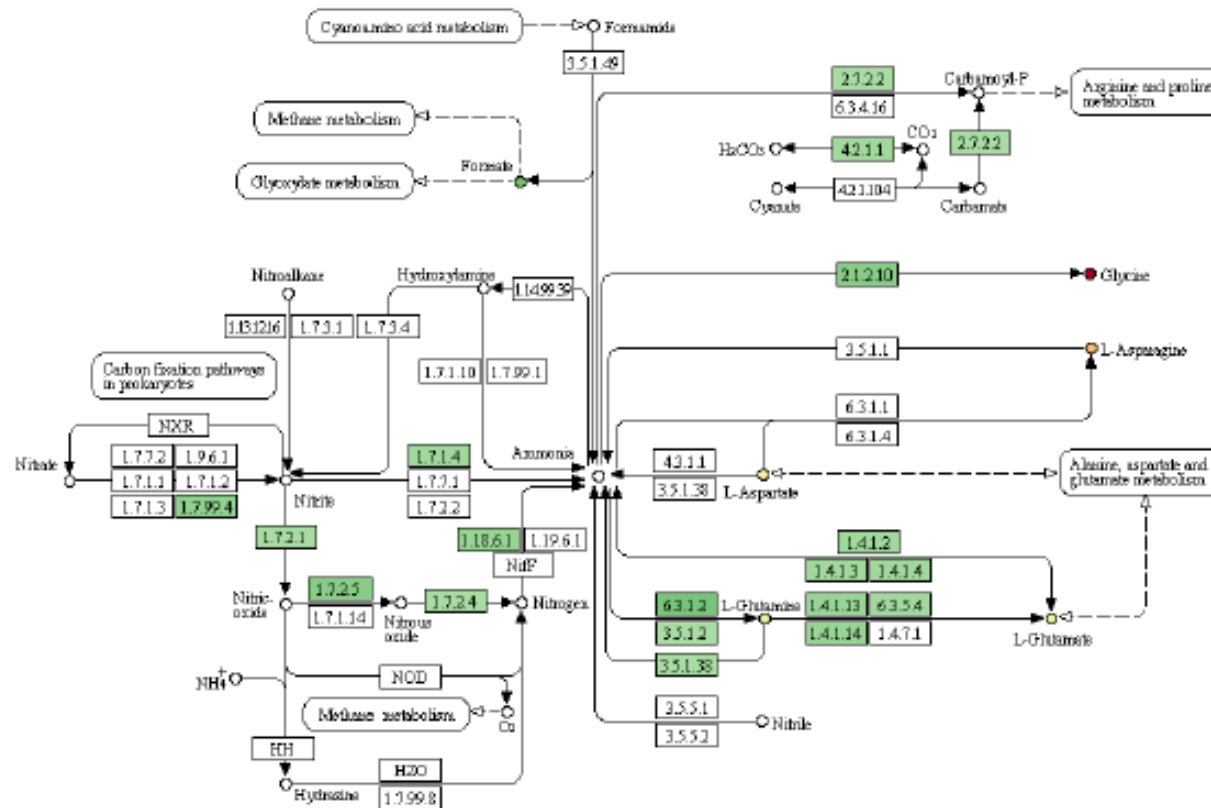


**AV rings:** overall strains comparison

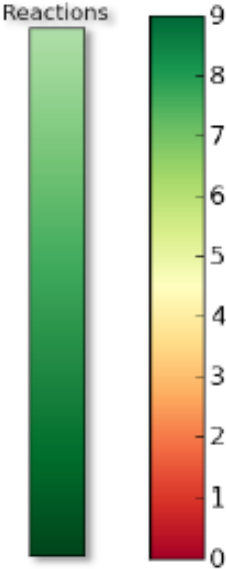


# dape: single genome metabolic network

## Nitrogen metabolism



Reactions



### Interactive metabolic maps (as web pages)

- Reactions copy number
- Compounds AV



**Rm1021**

Nitrate

Carbon fixation pathways in prokaryotes

NXR

1.7.7.2	1.9.6.1
1.7.1.1	1.7.1.2
1.7.1.3	1.7.99.4

Nitrite

Nitroalkane

Hydroxylamine

Ammonia

Nitric-oxide

Nitrous oxide

Nitrogen

NifF

1.7.2.1	1.7.2.5	1.7.1.14	1.7.2.4
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**AK58**

Nitrate

Carbon fixation pathways in prokaryotes

NXR

1.7.7.2	1.9.6.1
1.7.1.1	1.7.1.2
1.7.1.3	1.7.99.4

Nitrite

Nitroalkane

Hydroxylamine

Ammonia

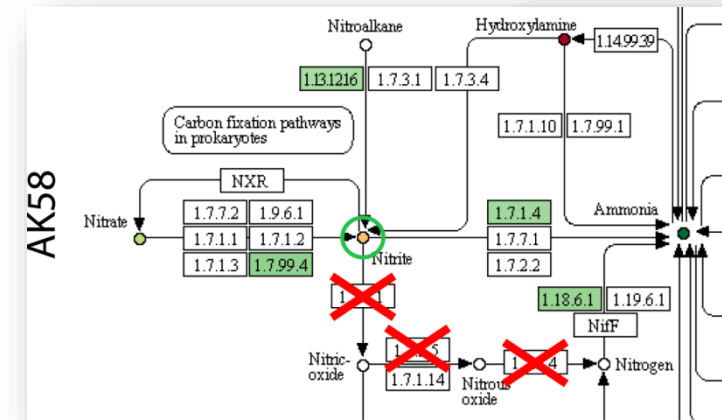
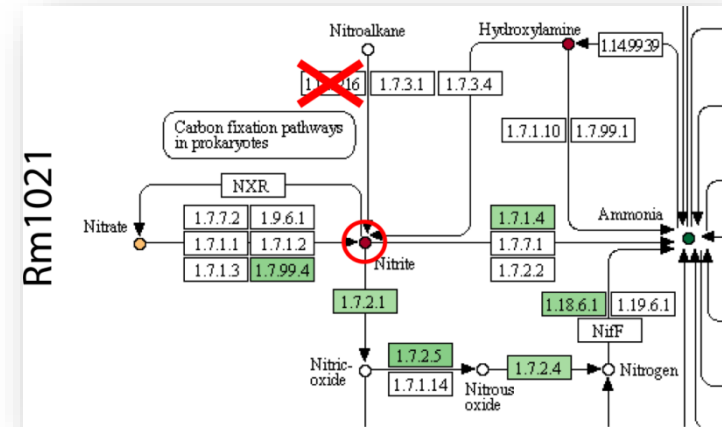
Nitric-oxide

Nitrous oxide

Nitrogen

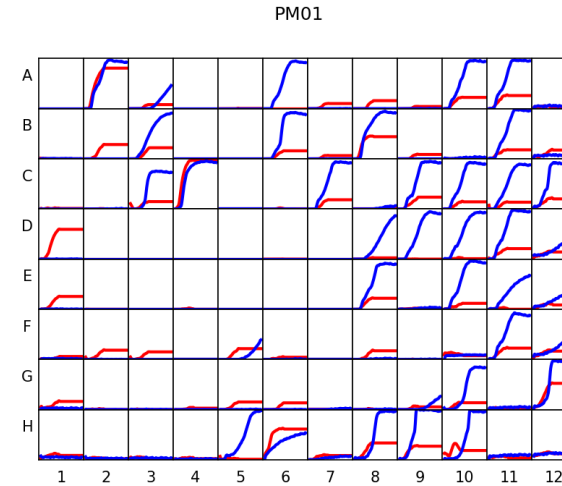
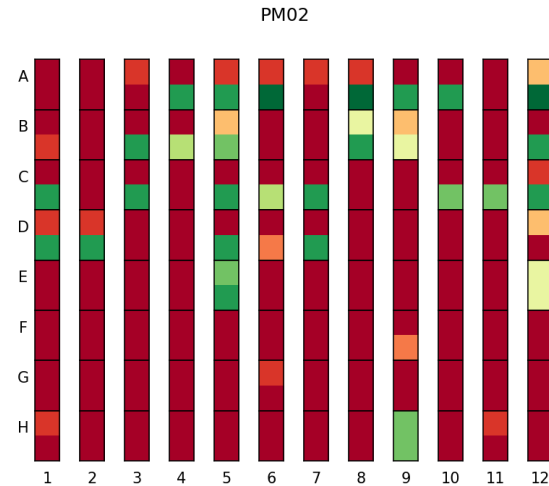
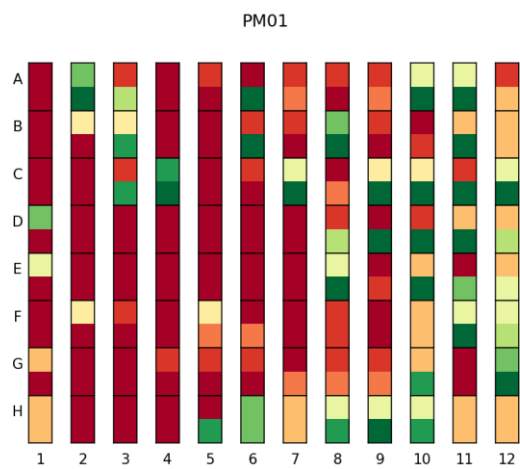
NifF

1.13.12.6	1.7.3.1	1.7.3.4
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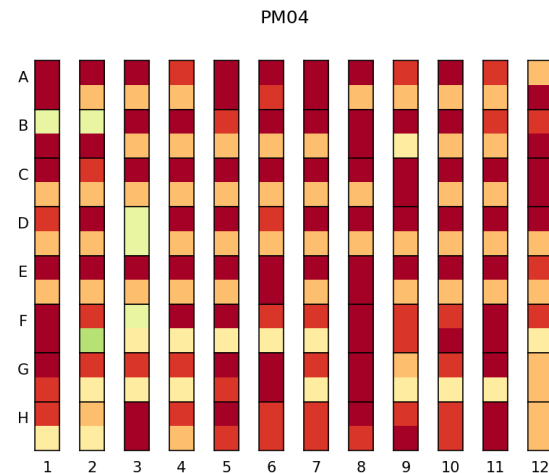
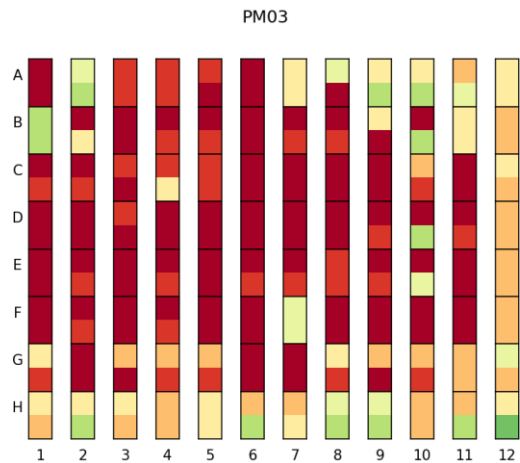
# Case study n.1

## Bacillus subtilis vs Paenibacillus polymixa (PM1,2,3 e 4)



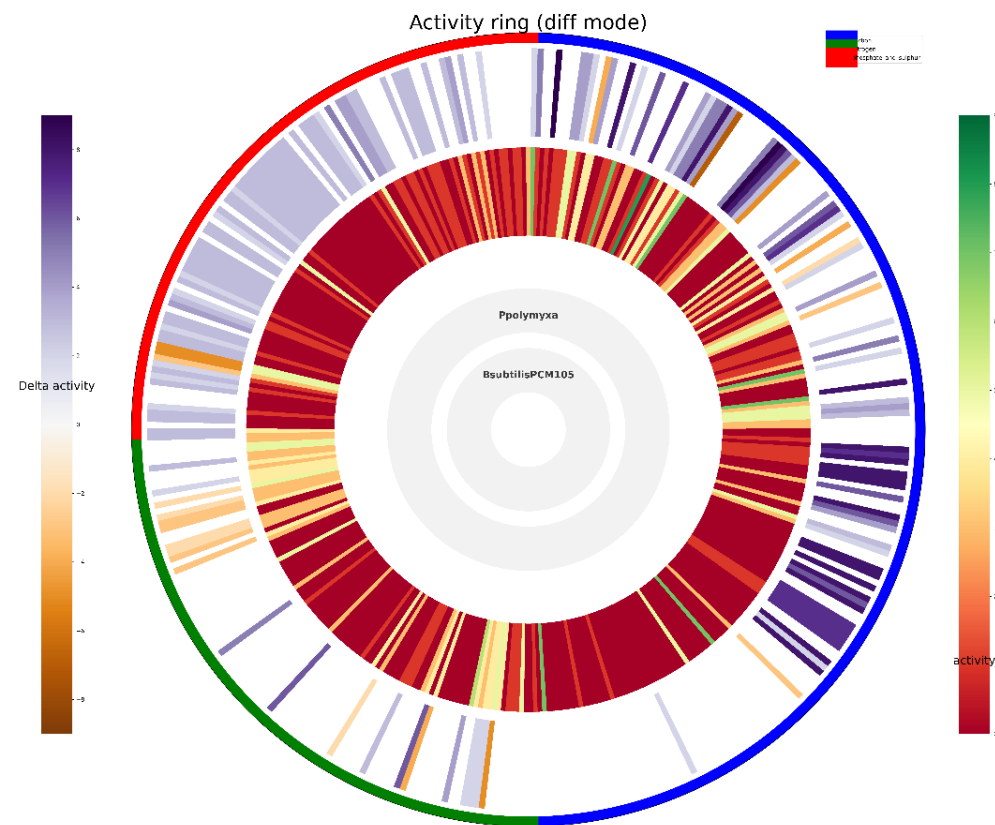
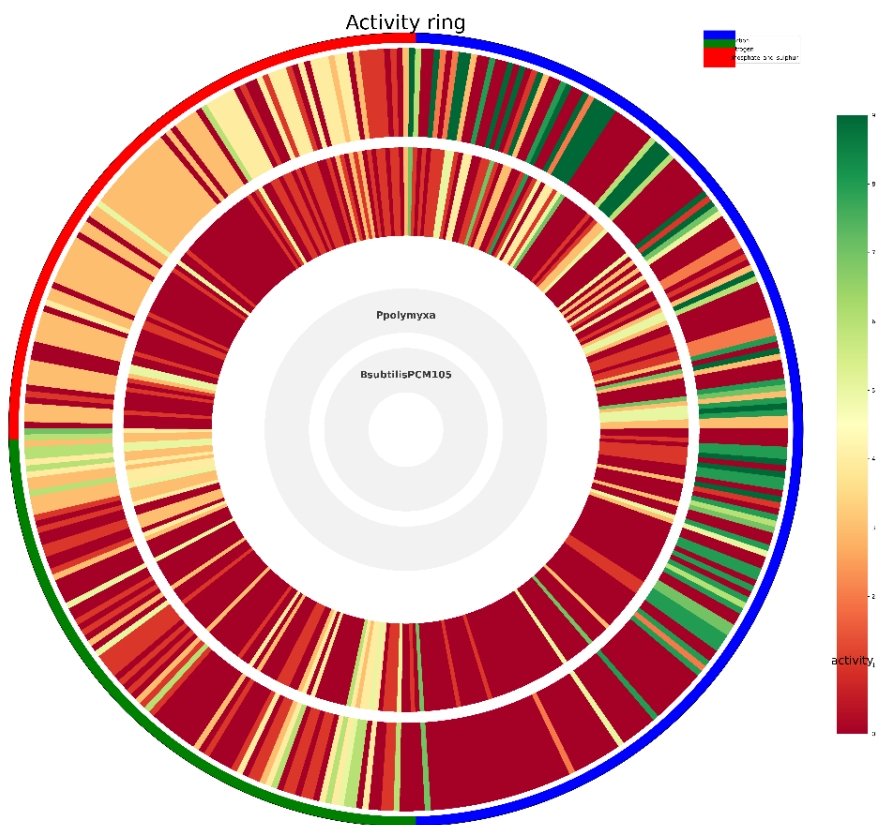
Strains color codes and order in heatmap (PM01)

<b>BsubtilisPCM105</b>
<b>Ppolymixa</b>



# Case study n.1

Bacillus subtilis vs Paenibacillus polymixa (PM1,2,3 e 4)





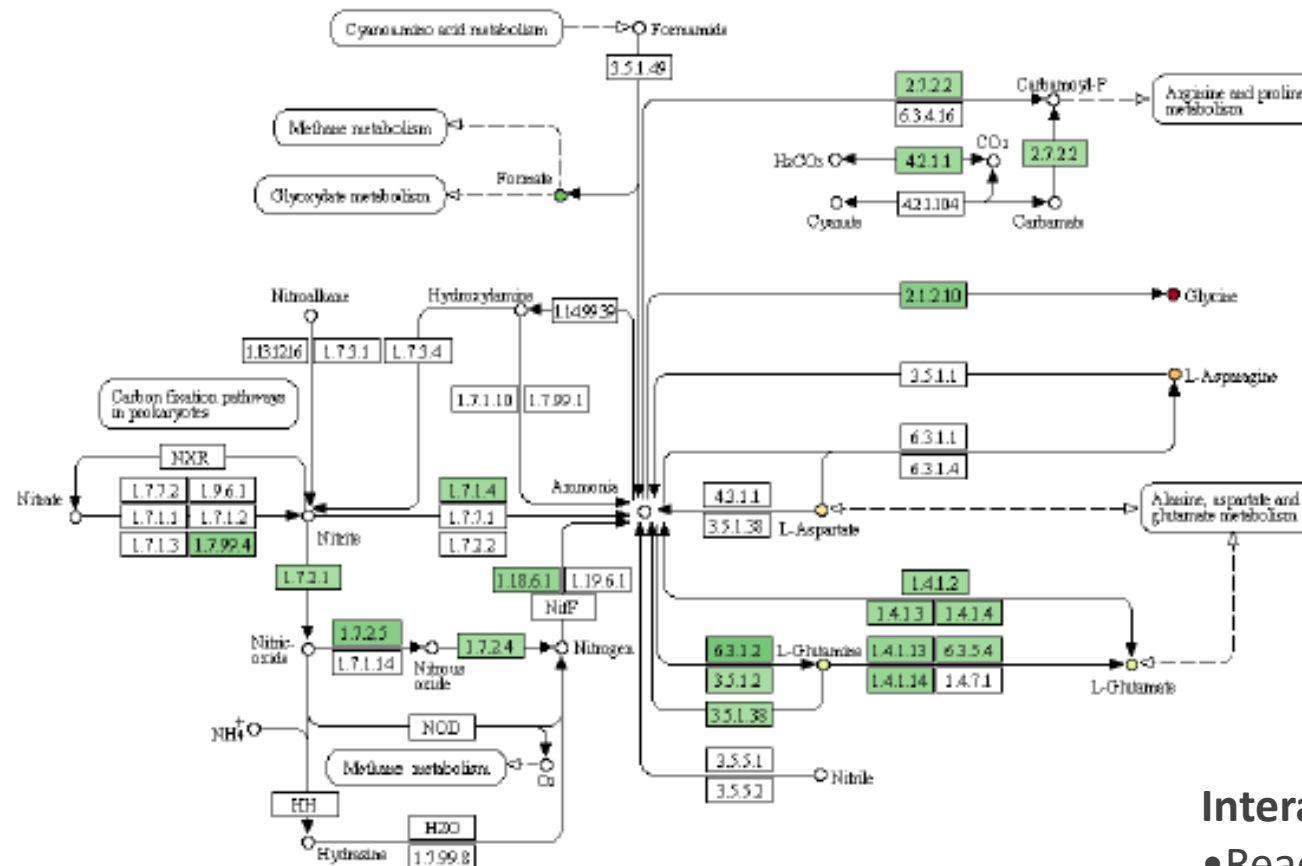
# Case study n.1

Bacillus subtilis vs Paenibacillus polymixa (PM1,2,3 e 4)

Pathway	Name	Category	BsubtilisPCM105		Ppolymixa	
			Reactions	meanAV	Reactions	meanAV
path:map01240	Biosynthesis of cofactors	nitrogen	23,00	1,82	35,00	3,36
path:map01240	Biosynthesis of cofactors	carbon	23,00	1,75	35,00	1,88
path:map01240	Biosynthesis of cofactors	phosphate_and_sulphur	23,00	0,33	35,00	2,83
path:map00230	Purine metabolism	carbon	21,00	2,29	20,00	3,00
path:map00230	Purine metabolism	nitrogen	21,00	2,25	20,00	2,17
path:map00230	Purine metabolism	phosphate_and_sulphur	21,00	0,88	20,00	2,75
path:map01232	Nucleotide metabolism	carbon	15,00	3,50	16,00	5,50
path:map01232	Nucleotide metabolism	nitrogen	15,00	1,29	16,00	1,21
path:map01232	Nucleotide metabolism	phosphate_and_sulphur	15,00	0,00	16,00	3,00
path:map00061	Fatty acid biosynthesis	carbon	15,00	0,00	15,00	0,00
path:map00061	Fatty acid biosynthesis	phosphate_and_sulphur	15,00	NA	15,00	NA
path:map00061	Fatty acid biosynthesis	nitrogen	15,00	NA	15,00	NA
path:map00640	Propanoate metabolism	carbon	9,00	0,20	6,00	0,00
path:map00640	Propanoate metabolism	phosphate_and_sulphur	9,00	NA	6,00	NA
path:map00640	Propanoate metabolism	nitrogen	9,00	NA	6,00	NA
path:map00620	Pyruvate metabolism	carbon	8,00	2,83	6,00	3,67
path:map00620	Pyruvate metabolism	phosphate_and_sulphur	8,00	0,00	6,00	3,00
path:map00620	Pyruvate metabolism	nitrogen	8,00	NA	6,00	NA
path:map00240	Pyrimidine metabolism	carbon	7,00	3,25	10,00	3,75
path:map00330	Arginine and proline metabolism	nitrogen	7,00	3,11	4,00	1,67

# Case study n.1

KEGG Database: Nitrogen metabolism (map1240)

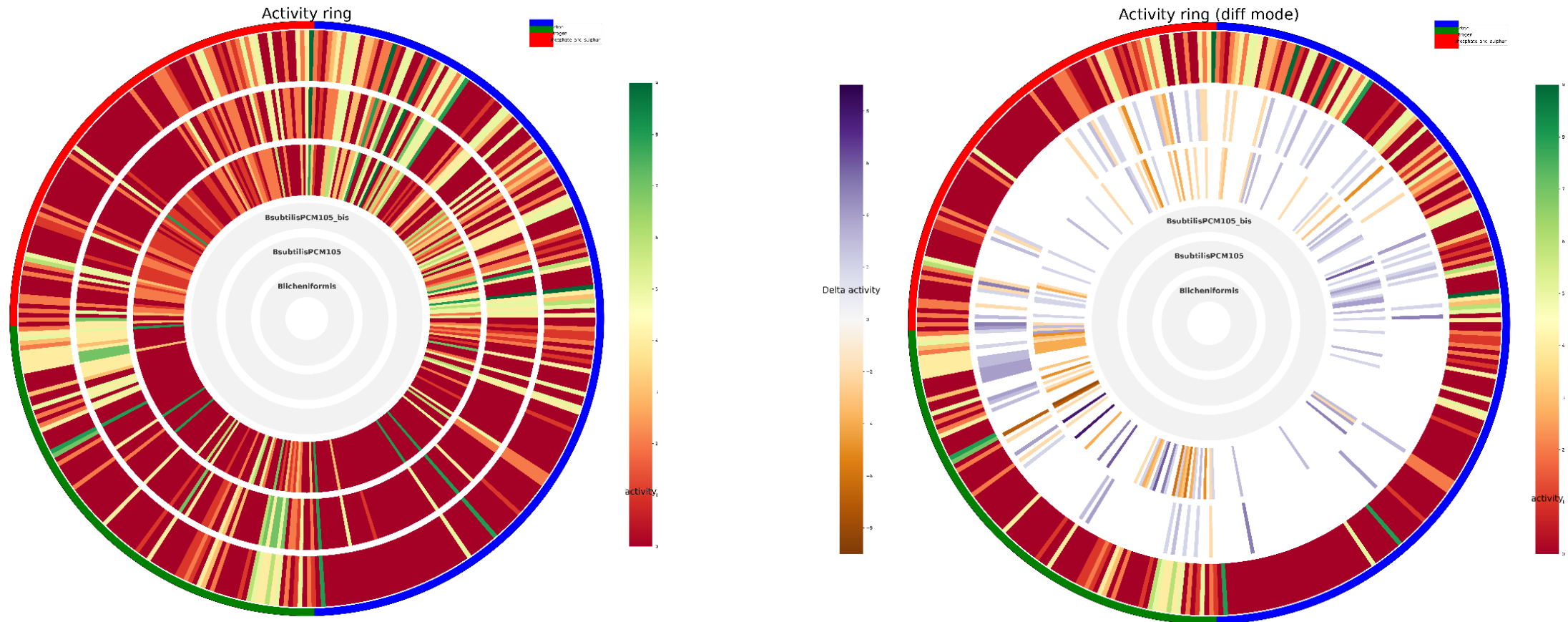


Interactive metabolic maps (as web pages)

- Reactions copy number
- Compounds AV

# Case study n.2

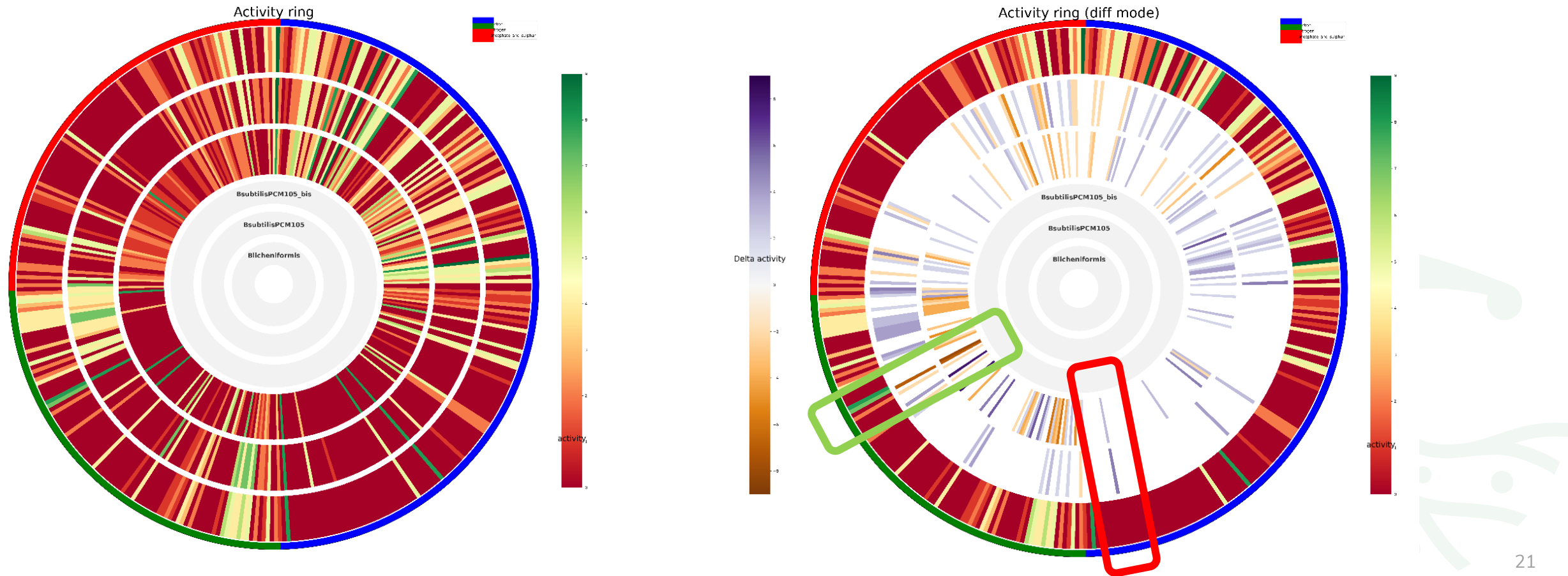
Bacillus subtilis and B.licheniformis (PM 1,2,3 e 4), single strains and co-inoculated





# Case study n.2

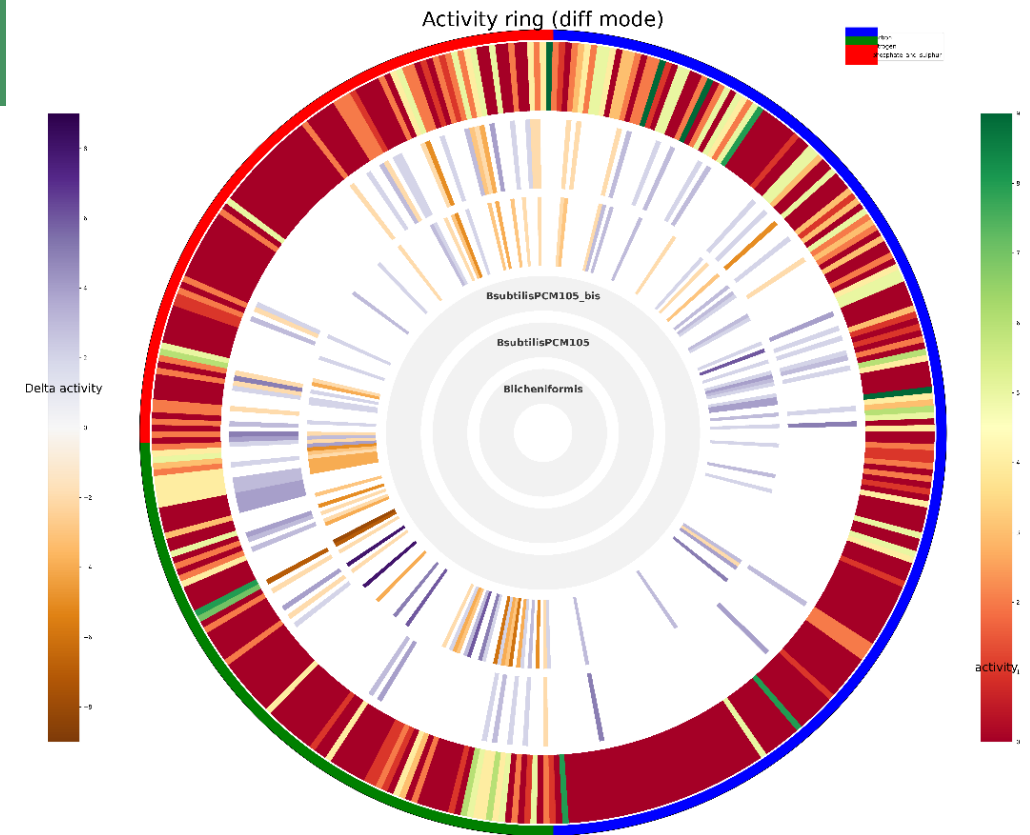
Bacillus subtilis and B.licheniformis (PM 1,2,3 e 4), single strains and co-inoculated



# Case study n.2

Bacillus subtilis and B.licheniformis (PM 1,2,3 e 4)

From AV data (diff mode) you can easily identify the substrates that are metabolized (or not) in the co-inoculum as compared to the single strains



category	moa	co_id	Blicheniformis	BsubtilisPCM105	Mean activity	Colnoculum	Difference	ass(diff)
nitrogen	N-Source, other	C00242	0	0	0	7	7	7
phosphate & sulphur	S-Source, organic		0	0	0	4	4	4
carbon	C-Source, carbohydrate	C00124	0	0	0	2	2	2
carbon	C-Source, carbohydrate	C01019	0	0	0	2	2	2
carbon	C-Source, alcohol	C00466	0	0	0	2	2	2
nitrogen	N-Source, other	C03570	0	0	0	2	2	2
phosphate & sulphur	P-Source, organic		0	0	0	2	2	2
phosphate & sulphur	S-Source, organic	C06809	0	0	0	2	2	2
phosphate & sulphur	S-Source, organic	C11145	0	0	0	2	2	2

# Case study n.2

## GC/MS measurements (Volatile Organic Compounds)

	Class	Bio-compound
	Acetates	Propyl acetate
	Alcohols	2-Methyl-1-propanol
	Alcohols	3-methyl-1-Butanol
	Alcohols	2-methyl-1-Butanol
	Alcohols	1-Propanol
	Alcohols	1-Butanol
	Alcohols	Isopropyl Alcohol
	Alcohols	Dimethylsilanediol
	Alkanes	3-Methylbutyronitrile
	Alkanes	Dimethyl disulfide
	Alkanes	1-isothiocyanato-2-methylpropane
	Alkanes	2,3,4-Trimethyloxetane
	Alkanes	2-methyl-3-propan-2-yloxirane
	Alkanes	Pentane
	Alkanes	1-Chlorobutane
	Alkanes	Dodecamethylcyclohexasiloxane
	Alkanes	Decamethylcyclopentasiloxane
	Alkanes	Hexamethylcyclotrisiloxane
	Alkanes	2,4-Dimethylhexane
	Alkanes	Isobutane
	Alkanes	3,4,5,6-Tetramethyloctane
	Alkanes	1-Chloropentane
	Alkanes	Tetradecamethylcycloheptasiloxane
	Alkanes	Octamethyltetrasiloxane
	Alkenes	2-methyl-1,3-butadiene
	Alkenes	1,4-Pentadiene
	Alkenes	Petasitene
	Alkenes	3-Ethyl-1-cyclopentene
	Alkenes	1,3-Pentadiene
	Amides	Formamide
	Amines	Ethylenimine
	Arene	Benzene
	Esters	2-Methylpropanoic acid ethyl ester
	Esters	2-Methylbutanoic acid ethyl ester
	Esters	3,5-difluorophenyl 2-(trifluoromethyl)benzoate
	Esters	Ethyl 3-methylbutanoate
	Esters	Succinic acid, bis(trimethylsilyl) ester
	Esters	Propyl 2-methylbutanoate
	Esters	Propyl isobutyrate
	Esters	Propyl propanoate
	Ethers	Dimethyl ether
	Ethers	Ethyl-1-propenyl ether
	Ethers	Propyl methyl ether
	Ketones	2-Butanone
	Ketones	Methoxy-2-propanone
	Ketones	3-Methyl-2-pentanone
	Ketones	2-Pentanone
	Ketones	2-Nonanone
	Ketones	Acetoin
	Ketones	Acetone
	Ketones	2-Hexanone
	Ketones	6-Methyl-3-heptanone
	Ketones	1-Butenyl methyl ketone
	Ketones	1-Methyl-4-acetylcyclohex-1-ene
	Ketones	Methyl 1-cyclohexenyl ketone
	Phenols	3,4,5-Trimethylphenol
	Pyrazines	Pyrazine, 2-(1-methylethyl)-5-(2-methylpropyl)-
	Pyrazines	Pyrazine, 2,3-dimethyl-5-(1-methylpropyl)-
	Pyrazines	3-iso propyl-2,5-dimethylpyrazine
	Pyrazines	3-Isopropyl-2-methylpyrazine
	Pyrazines	1,2,4-Triazolo[4,3-a]pyridine
	Pyrazines	2,3-Dimethyl-5-sec-butylpyrazine
	Pyrazines	3,5-Dimethyl-2-propylpyrazine
	Pyrazines	2,4,6-Trimethylpyridine
<i>KIS_Bacillus cf firmus</i>		
<i>Stenotrophomonas rhizophilla</i> SPA P69		
<i>Serratia plymutica</i> HRO-C48		
<i>Serratia plymutica</i> HRO-C48_		
<i>Stenotrophomonas rhizophilla</i> SPA P69		
INHORT_Paenibacillus K16		
INTERMAG_Bacillus licheniformis		
INTERMAG_Bacillus subtilis		
INTERMAG_Bacillus licheniformis_Bacillus subtilis		

# Case study n.2

## GC/MS measurements (Volatile Organic Compounds)

	Class	Bio-compound
	Acetates	Propyl acetate
	Alcohols	2-Methyl-1-propanol
	Alcohols	3-methyl-1-Butanol
	Alcohols	2-methyl-1-Butanol
	Alcohols	1-Propanol
	Alcohols	1-Butanol
	Alcohols	Isopropyl Alcohol
	Alcohols	Dimethylsilanediol
	Alkanes	3-Methylbutyronitrile
	Alkanes	Dimethyl disulfide
	Alkanes	1-isothiocyanato-2-methylpropane
	Alkanes	2,3,4-Trimethyloxetane
	Alkanes	2-methyl-3-propan-2-yloxirane
	Alkanes	Pentane
	Alkanes	1-Chlorobutane
	Alkanes	Dodecamethylcyclohexasiloxane
	Alkanes	Decamethylcyclopentasiloxane
	Alkanes	Hexamethylcyclotrisiloxane
	Alkanes	2,4-Dimethylhexane
	Alkanes	Isobutane
	Alkanes	3,4,5,6-Tetramethyloctane
	Alkanes	1-Chloropentane
	Alkanes	Tetradecamethylcycloheptasiloxane
	Alkanes	Octamethyltetrasiloxane
	Alkenes	2-methyl-1,3-butadiene
	Alkenes	1,4-Pentadiene
	Alkenes	Petasitene
	Alkenes	3-Ethyl-1-cyclopentene
	Alkenes	1,3-Pentadiene
	Amides	Formamide
	Amines	Ethylenimine
	Arene	Benzene
	Esters	2-Methylpropanoic acid ethyl ester
	Esters	2-Methylbutanoic acid ethyl ester
	Esters	3,5-difluorophenyl 2-(trifluoromethyl)benzoate
	Esters	Ethyl 3-methylbutanoate
	Esters	Succinic acid, bis(trimethylsilyl) ester
	Esters	Propyl 2-methylbutanoate
	Esters	Propyl isobutyrate
	Esters	Propyl propanoate
	Ethers	Dimethyl ether
	Ethers	Ethyl-1-propenyl ether
	Ethers	Propyl methyl ether
	Ketones	2-Butanone
	Ketones	Methoxy-2-propanone
	Ketones	3-Methyl-2-pentanone
	Ketones	2-Pentanone
	Ketones	2-Nonanone
	Ketones	Acetoin
	Ketones	Acetone
	Ketones	2-Hexanone
	Ketones	6-Methyl-3-heptanone
	Ketones	1-Butenyl methyl ketone
	Ketones	1-Methyl-4-acetylcyclohex-1-ene
	Ketones	Methyl 1-cyclohexenyl ketone
	Phenols	3,4,5-Trimethylphenol
	Pyrazines	Pyrazine, 2-(1-methylethyl)-5-(2-methylpropyl)-
	Pyrazines	Pyrazine, 2,3-dimethyl-5-(1-methylpropyl)-
	Pyrazines	3-iso propyl-2,5-dimethylpyrazine
	Pyrazines	3-Isopropyl-2-methylpyrazine
	Pyrazines	1,2,4-Triazolo[4,3-a]pyridine
	Pyrazines	2,3-Dimethyl-5-sec-butylpyrazine
	Pyrazines	3,5-Dimethyl-2-propylpyrazine
	Pyrazines	2,4,6-Trimethylpyridine
KIS_ <i>Bacillus cf firmus</i>		
Stenotrophomonas rhizophilla SPA P69		
Serratia plymutica HRO-C48		
Serratia plymutica HRO-C48_ Stenotrophomonas rhizophilla SPA P69		
INTERMAG_ <i>Bacillus licheniformis</i>		
INTERMAG_ <i>Bacillus subtilis</i>		
INTERMAG_ <i>Bacillus licheniformis</i> _ <i>Bacillus subtilis</i>		



# Conclusions

- The BIOLOG technology strongly support the genomics analysis in a complementary way
- Ductape software includes the tools to analyze genomes and phenomes as well as linking them with specific metabolic pathways
- More specifically, the definition of a AV and the visualization options included, offer the possibility to successfully and easily compare multiple strains together
- More specific details on data analysis will be presented in the next presentation



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**Thanks for you attention**