Big Data Analytics and Molecular Diagnostic Methods for Treatment Pathways of Priority Pathogens



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WHO Priority Pathogens List (PPL) - 2017

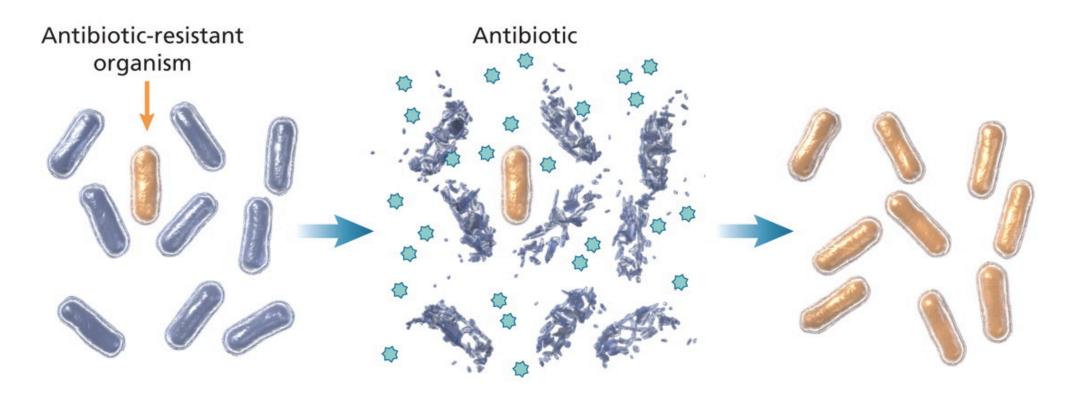
CRITICAL	HIGH	MEDIUM		
Acinetobacter baumannii, carbapenem- resistant	Enterococcus faecium, vancomycin-resistant	Streptococcus pneumoniae, penicillin-non-susceptible		
Pseudomonas aeruginosa, carbapenem- resistant	Staphylococcus aureus, methicillin-resistant, vancomycin intermediate resistant	Haemophilus influenzae, ampicillin- resistant		
Enterobacteriaceae*, carbapenem-resistant, 3 rd generation cephalosporin-resistant	Neisseria gonorrhoeae, 3 rd generation cephalosporin-resistant, fluoroquinoloneresistant	Shigella, fluoroquinolone-resistant		
* Enterobacteriaceae include: Klebsiella	Campylobacter, fluoroquinolone-resistant			
pneumonia, Escherichia coli, Enterobacter spp., Serratia spp.,	Salmonella, fluoroquinolone-resistant	ACCORDING TO THE STUDY, MTB IS ON THE PRIORITY FOR NEW DRUG DISCOVERY		
Proteus spp., and Providencia spp, Morganella spp	Helicobacter pylori, clarithromycin-resistant	AND DEVELOPMENT		

Antimicrobial resistance occurs when microorganisms such as bacteria, viruses, fungi and parasites change in ways that render the medications used to cure the infections they cause ineffective.

When the microorganisms become resistant to most antimicrobials they are often referred to as "superbugs".



Antibiotic Resistance



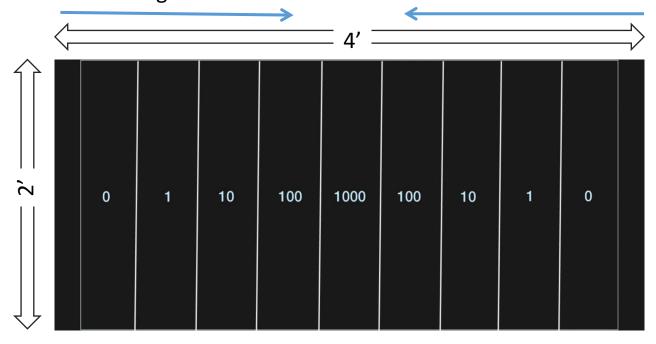
Population of bacteria with a subset of antibiotic-resistant organisms.

In the presence of an antibiotic, susceptible strains are killed; the resistant strain survives.

The resistant strain proliferates and may be capable of causing a new infection.

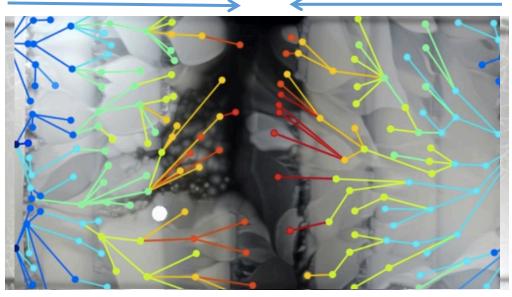
A cinematic approach to drug resistance – Harvard Gazette

Increasing concentration of antibiotics towards the center



An experimental device for studying microbial evolution in a spatially structured environment

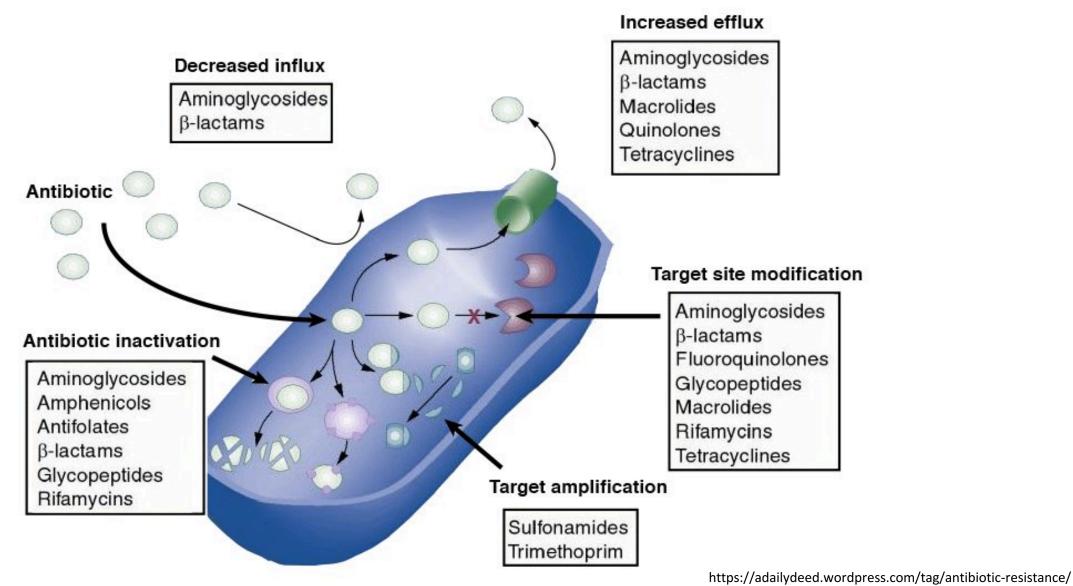
E. coli appear as white on the black background



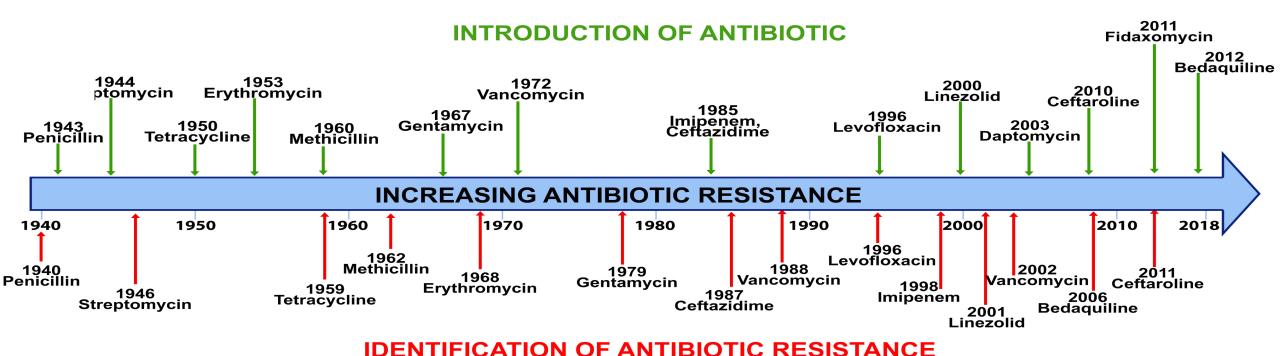
Bacteria spread as they evolve

Microbial Evolution and Growth Arena (MEGA) plate

Drug Resistance Mechanisms



Timeline showing discovery of antibiotics, year of introduction and observation of resistance

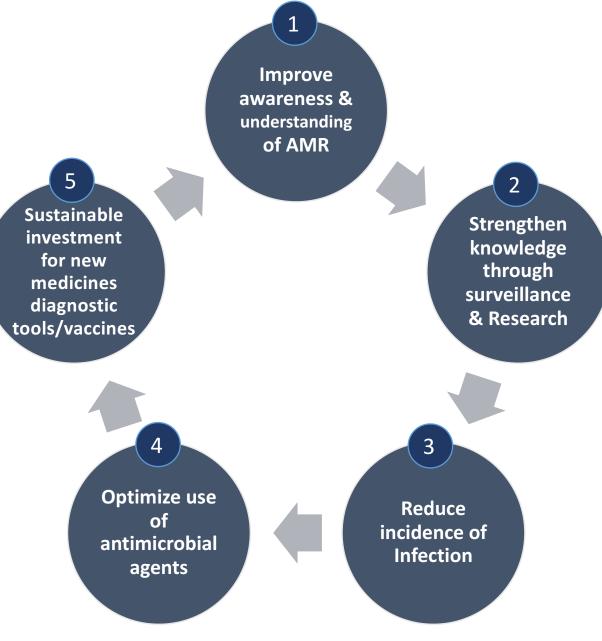


- > Discovery of antibiotics is rapidly followed by emergence of resistance against them.
- Continuing improvements are being made for discovery of new antimicrobial agents but are outpaced by emerging and re-emerging drug resistant infectious disease agents.

© CSIR-IMTECH



Sixty-eight World Health Assembly in May 2015 endorsed a global action plan to tackle antimicrobial resistance



Culture remains the gold standard for diagnosis of the Pathogen So what is wrong?

- As culture takes time empirical based approach is used for diagnosis and prescription (broad-spectrum antibiotics are prescribed to cover as many pathogens)
- Leads to overtreatment or not the right treatment
- Impacts the emergence of antimicrobial resistance
- <60% sensitive; biological sample has low number of pathogens
- Complex etiology of conditions like sepsis makes it even more challenging 90–95% of blood cultures remain negative in immunocompromised patients suspected of infection; even in the cases where bacterial or fungal sepsis is likely

	No.	% Total	Sites of Infection
Lung	1016	37.2	
Intraabdominal	801	29.3	
Bowel perforation/peritonitis	226	8.3	
Postoperative bowel perforation/anastomotic dehiscence	65	2.4	
Spontaneous bacterial peritonitis	50	1.8	
Other peritonitis	18	0.7	
Intraabdominal abscess	44	1.6	
Cholecystitis	40	1.5	
Ascending cholangitis	43	1.6	
Ischemic bowel/bowel infarction	166	6.1	
Clostridium difficile enterocolitis/toxic	47	1.7	
megacolon			Concic is a systematic
Genitourinary	293	10.7	Sepsis is a systematic
Skin and soft tissue	197	7.2	inflammatory response
Necrotizing soft tissue infections	74	2.7	•
Cellulitis	46	1.7	caused by infection
Operative wound infection	22	0.8	, , , , , , , , , , , , ,
Soft tissue abscess	20	0.7	
Decubitus ulcer	16	0.6	DaNIC Ctudy
Diabetic lower extremity ulcer/cellulitis	13	0.5	DeNIS Study
Surgical site infection	31	1.1	
Central nervous system infection (meningitis/	20	0.7	
abscess)			
Intravascular catheter infection	100	3.7	
Primary bloodstream infection (bacteremia	120	4.4	
without identifiable source)	120		
Systemically disseminated infection (including	58	2.1	
	30	2.1	
yeast and tuberculosis)	21	0.0	
Septic arthritis Mediastinitis	15	0.8	
		0.5	
Other	59	2.1	

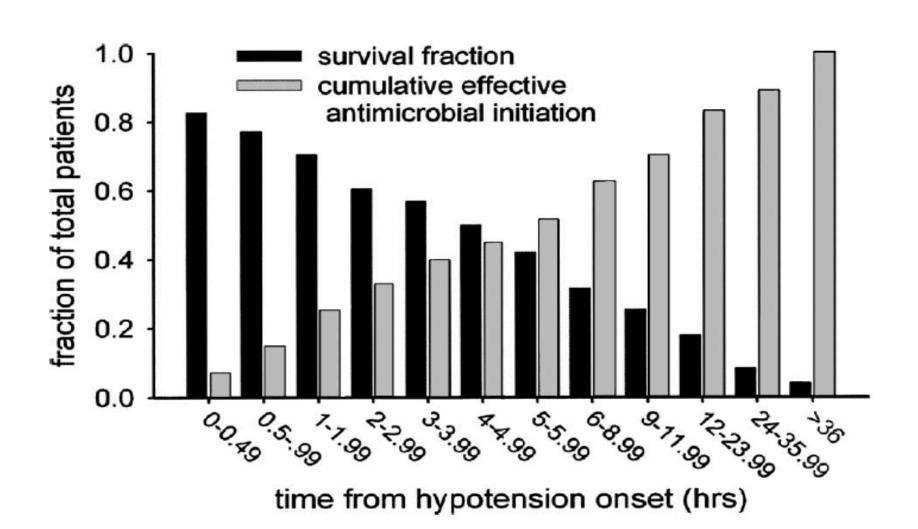


Pathogen	No. of Patients	% Total
Gram-negative organisms	930	47.9
Escherichia coli	435	22.4
Klebsiella species	131	6.7
Pseudomonas aeruginosa	115	5.9
Enterobacter species	80	4.1
Haemophilus influenzae	44	2.2
Proteus species	25	1.2
Acinetobacter species	21	1.1
Serratia species	20 16	1.0 0.8
Stenotrophomonas maltophila		0.8
Morganella morganii	14 13	0.7
Citrobacter species Neisseria meningitidis	6	0.7
Burkholderia cepacia	3	0.3
Haemophilus parainfluenzae	3	0.2
Other Gram-negative bacilli	8	0.4
Gram-positive organisms	731	38.3
Staphylococcus aureus	302	15.6
Streptococcus pneumoniae	170	8.8
Streptococcus faecalis	77	4.0
Group A Streptococcus species	69	3.6
Other B-hemolytic streptococci	43	2.2
Viridans streptococci	37	1.9
Streptococcus faecium	29	1.5
Bacillus species	5	0.3
Corynebacterium jeijkeium	5	0.3
Staphylococcus lugdunensis	1	0.3
Yeast/fungi	160	8.2
Candida albicans	91	4.7
Candida glabrata	18	0.9
Aspergillus/Mucor species	14	0.7
Blastomyces species	10	0.5
Candida tropicalis	4	0.2
Candida parapsilosis	4	0.2
Candida krusei	3	0.2
Cryptococcus neoformans	1	0.1
Histoplasma species	1	0.1
Other unidentified yeast	13	0.6
Anaerobes Clostridium difficile	69 46	3.6 2.4
Bacteroides fragilis	46 15	0.8
Other clostridia	8	0.4
Legionella species	8	0.4
Mycobacterium tuberculosis	11	0.4
rigeoodeteriam taberealosis	11	0.0

Suspected primary pathogens in septic shock



The Need for Rapid Diagnosis

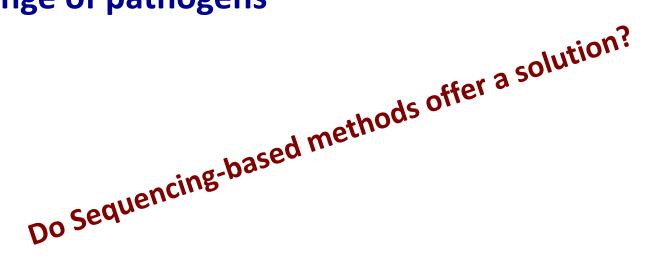


https://pubmed.ncbi.nlm.nih.gov/16625125/



Need a method that:

- Non culturable pathogen detection
- Can detect an entire range of pathogens
- Comprehensive
- Sensitive
- Specific
- Rapid
- Works with dead or live pathogen
- Differentiate between pathogen and human cells



What sequencing based methods can offer?

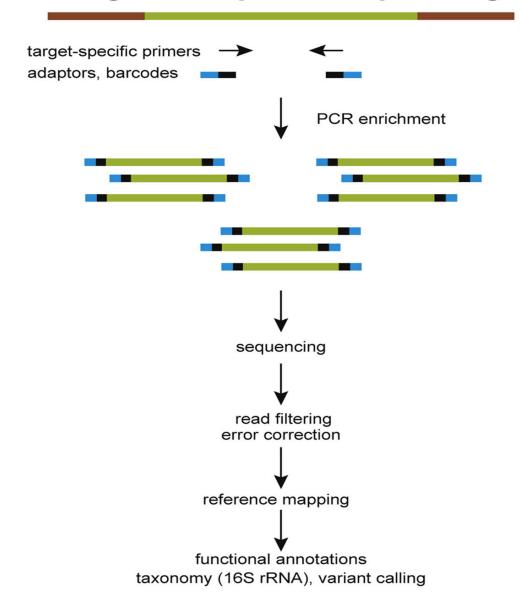
- Identify unculturable pathogens
- Distinguish between different pathogen strains
- Detect new drug resistance determinants, e.g. detection of mecC in S. aureus*
- Rapid with extreme accuracy, -> better-informed treatment decisions
- Sequence data can be used to develop specific primers for PCR-based assays
 - to provide more complete sequence information and
 - to develop other specific field tests specific genotype based

Traditional vs Next Generation Sequencing Platforms in Clinical Microbiology

- The Sanger method has limited sensitivity for minor variants when present at <15% to 20% of the viral population, whereas NGS methods can detect drug-resistant mutations (DRMs) present at approximately 1%
- At least half of the DRMs identified by NGS are missed by Sanger sequencing

 The presence of such variants has been shown to predict an increased risk for therapy failure

Targeted amplicon sequencing

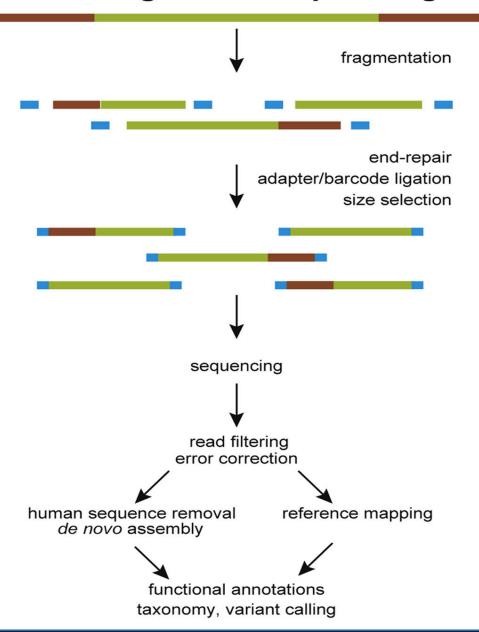


Cannot determine new pathogens or

New antimicrobial resistance genes

Not comprehensive

Whole-genome sequencing



Comprehensive

Can help in detecting novel AMR genes

Nanopore - Minlon



From bench to bedside – Point of care sequencing - promise

Companies which are pushing the boundaries of PoC with NGS in IDD

COMPANY	REGION	PRODUCT TYPE(S)	DESCRIPTION	FUNDING
Illumina and bioMérieux	U.S.	NGS Service	Illumina, the dominant player in the NGS market, and bioMérieux, a leader in <i>in vitro</i> diagnostics, partnered together to launch an NGS service dedicated to infectious disease diagnostics. ¹⁴	Through a partnership between Illumina and bioMérieux, both global players in healthcare. 15
Oxford Nanopore Technologies	U.K	 NGS Sequencers NGS Assays Sample Prep Kits Library Prep Kits 	Founded in 2005, the company is developing portable NGS sequencing products for the analysis of single molecules. ^{22, 27}	 Raised a total of \$386 million through VC funding. Closed a \$109 million round in July 2015 from new and existing investors in the U.K., U.S., and mainland Europe for product development, manufacturing and commercialization.²⁰
DNAe	U.K.	NGS SequencersNGS Assays	Founded in 2003, DNAe aims to revolutionize NGS diagnostic technology by bringing the entire NGS workflow from sample preparation, sequencing, and analysis into one semiconductor chip. ⁴	 Secured a \$38 million bank facility from Citibank in November 2015 for development of an NGS in vitro diagnostic test for Serious Blood Infections.⁵ Completed a Series A fundraising round in April 2014 for an undisclosed amount.⁵
MRIGlobal	U.S.	NGS Sequencers NGS Assays	Founded in 1977, MRIGlobal is an independent organization that performs contract research for the government and the industry.	Awarded \$14.8 million in February 2015 from the Defense Threat Reduction Agency of the U.S. Department of Defense to develop a comprehensive NGS platform for infectious diseases diagnosis.8
Pathoquest	France	NGS AssaysBioinformatics	Pathoquest is developing a comprehensive NGS assay for diagnosing infectious diseases, including bioinformatics.	Raised \$5 million in July 2013 in Series B funding by IDInvest Partners, Aurinvest and Kurma Partners. ²³

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Species	identiii	ican
From	clonal	sam

- NCBI BLAST
- GenBank
- · Other genome databases in Table 1

From nonclonal sample

Meta-assembly

- AMOS
- MIRA
- MetaVelvet

Clustering and species annotation

- MEGAN
- MG-RAST

Maximum likelihood phylogeny trees

- BEAST
- RAxML
- ClonalFrame
- ClonalOrigin

Whole-genome alignment

For SNP calling

- Mummer
- Mugsy

For structural variant calling

Mauve

Gene annotation

Bacterial

GLIMMER

RAST

Drug resistance in bacteria

- ResFinder
- ARG-ANNOT

A few databases and tools for DR determinant

Examples of Public Bioinformatics Databases That May Be Leveraged for Multiscale Analysis of Infectious Disease^a

S	Database Focus	For General Research	For Infectious Disease			
			Multipathogen	Pathogen-Specific		
	Genomes	 NCBI Nucleotide (GenBank/RefSeq) ENA/EMBL DDBJ 	ViPRNMPDRPATRICEuPathDB	 Influenza Research Database Tuberculosis Database LANL: Databases for HIV, HCV, and HFV 		
Clin	Gene products and functionality	UniProtKEGG	 Pathogen-Host Interaction Dat Antibiotic Resistance Genes E Comprehensive Antibiotic Resident Database 	Database		
in Infect Dis	Expression and immune profiles	GEOArrayExpress	• ImmPort			

List is not comprehensive



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Prediction of Susceptibility to First-Line Tuberculosis Drugs by DNA Sequencing

The CRyPTIC Consortium and the 100,000 Genomes Project

ABSTRACT

BACKGROUND

The World Health Organization recommends drug-susceptibility testing of Mycobacterium tuberculosis complex for all patients with tuberculosis to guide treatment deci-

The members of the writing group (Timothy M. Walker, D.Phil., A. Sarah Walker,



Lets know more about this study:

- 1. A total of 10,209 isolates were analyzed.
- 2. Resistance to isoniazid, rifampin, ethambutol, and pyrazinamide was correctly predicted with 97.1%, 97.5%, 94.6%, and 91.3% sensitivity, respectively
- 3. Susceptibility to these drugs was correctly predicted with 99.0%, 98.8%, 93.6%, and 96.8% specificity.
- 1. Among the 4037 phenotypic profiles that were predicted to be pansusceptible, 3952 (97.9%) were correctly predicted.

Susceptibility is predicted with higher accuracy

Clinical correlation of resistance phenotype is a limiting factor!!

Modes of action of first-and second-line anti-TB drugs, mechanisms of drug resistance and mutation frequency for each gene in clinical Mtb isolates

		•	_	
Group	Drug	Drug action	Drug resistance-associated gene(s)	Mutation frequency in clinical isolates (%) ^a
First-line anti-TB drugs	Rifampicin	Binding to the β-subunit of the RNA polymerase, inhibition of the elongation of messenger RNA	rpoB encoding for β-subunit of RNA polymerase	90-100
	Isoniazid	Activation by a catalase-peroxidase enzyme	katG encoding for catalase-peroxidase	40-97
		Inhibition of the synthesis of mycolic acids through binding to NADH-ACP-reductase	inhA encoding for fatty acid enoyl acyl carrier protein reductase A (InhA)	8-64
	Ethambutol	Inhibition of an arabinosyl transferase involved in cell wall synthesis	embB encoding for arabinosyl transferase	47-89
	Pyrazinamide	 Activation by the pyrazinamidase Disruption of membrane energetics that inhibits membrane transport 	pncA encoding for pyrazinamidase	44-97
	Streptomycin	Inhibition of protein synthesis by interaction with the	rrs encoding for 16S rRNA subunit	12-26
		16S rRNA and the S12 ribosomal protein	rpsL encoding for S12 ribosomal protein	40-68
		Inhibition of methylation of 16S rRNA	gidB encoding for 7-methylguanosine methyltransferase	5-13
Second-line anti-TB drugs	Amikacin, kanamycin, capreomycin	Inhibition of protein synthesis by interaction with the 16S rRNA	rrs encoding for 16S rRNA	40-90
	Kanamycin	Inhibition of acetyltransferase	eis encoding for aminoglycoside acetyltransferase	28-80
	Capreomycin	Inhibition of methylation of 16S rRNA & 23S rRNA	tlyA encoding for 2'-O-methyltransferase	4-13
	Ofloxacin, levofloxacin, moxifloxacin, gatifloxacin	Inhibition of the topoisomerase II (DNA gyrase) lead to the inhibition of DNA supercoiling	gyrA encoding for DNA gyrase subunit A and	70-90
			gyrB encoding for DNA gyrase subunit B	0-11
	Ethionamide	Inhibition of the synthesis of mycolic acids by interaction with NAD that inhibits the enoyl-ACP reductase	inhA encoding for fatty acid enoyl acyl carrier protein reductase A (InhA)	33-62
		Inhibition of metabolic activation by interaction with	ethA encoding for EthA	46-72
		the transcriptional repressor of the Monooxygenase (EthA)	ethR encoding for transcriptional repressor EthR, NADH-ACP	0-4





PMCID: PMC1216834

PMID: 16172379

Proc Natl Acad Sci U S A. 2005 Sep 27; 102(39): 13950–13955. Published online 2005 Sep 19. doi: 10.1073/pnas.0506758102 Genetics

Genome analysis of multiple pathogenic isolates of *Streptococcus* agalactiae: Implications for the microbial "pan-genome"

Hervé Tettelin, a,b Vega Masignani, b,c Michael J. Cieslewicz, b,d,e Claudio Donati,c Duccio Medini,c Naomi L. Ward, a,f Samuel V. Angiuoli, Jonathan Crabtree, Amanda L. Jones, A. Scott Durkin, Scott Durkin, Amanda L. Jones, A. Scott Durkin, Amanda L. Jones, A. Scott Durkin, Amanda L. Jones, Margarit y Ros, Company D. Peterson, Amanda L. Jones, Amanda L. Jones, Margarit y Ros, Company D. Peterson, Amanda L. Jones, Amanda L. Jon

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Strains of a bacterial species might differ substantially in their gene content and total gene pool of a species might be orders of magnitudes larger than the gene content of any single strain



The core genes are responsible for basic aspects of the biology of the species and its major phenotypic traits;

The accessory genes and singletons usually pertain to supplementary biochemical pathways and functions that may confer selective advantages such as ecological adaptation, virulence mechanisms, antibiotic resistance, or colonization of a new host.

n=30



Article | OPEN | Published: 17 October 2018

Machine learning and structural analysis of Mycobacterium tuberculosis pangenome identifies genetic signatures of antibiotic resistance

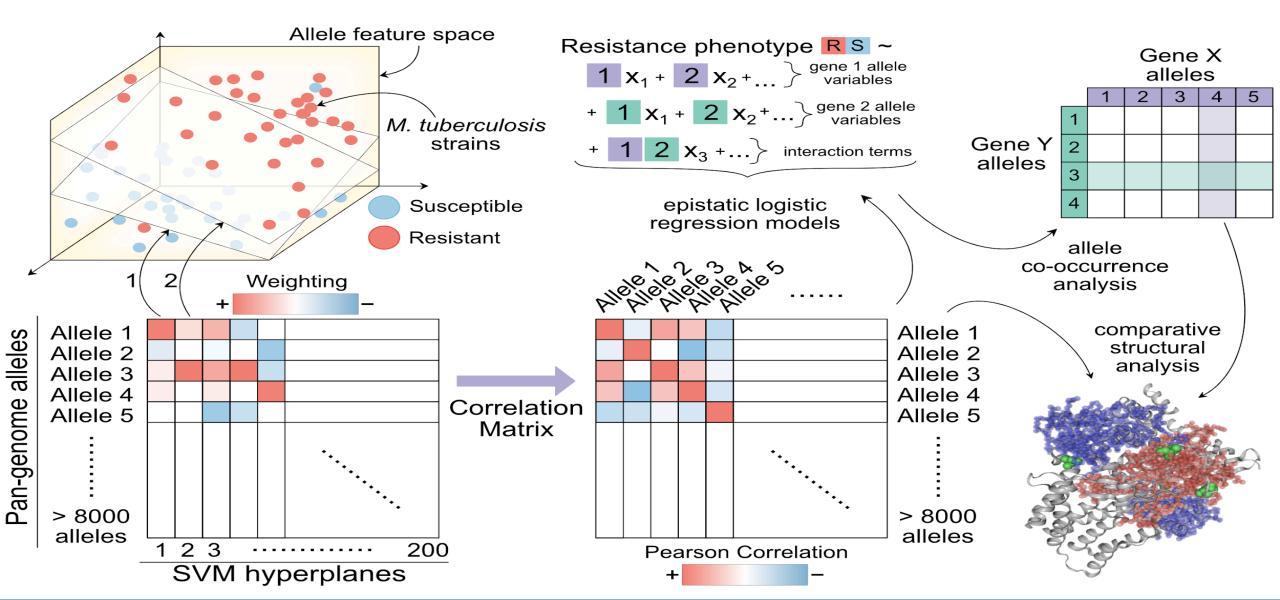
Erol S. Kavvas, Edward Catoiu, Nathan Mih, James T. Yurkovich, Yara Seif, Nicholas Dillon, David Heckmann, Amitesh Anand, Laurence Yang, Victor Nizet, Jonathan M. Monk № & Bernhard O. Palsson ▶



The dataset – Genome IDs with Drug Sensitivity Profiling

1	genome_id	isoniazid	rifampicin	ethambutol	amikacin	ciprofloxacin	clofazimine	cycloserine	ethionamide	kanamycin	moxifloxacin	nicotinamide
2	1295764_3	R	R	R			S		R	R		R
3	1423468_3	R	R	S								
4	1448744_3	R	R	S								
5	1448649_3	S	s	s								
6	1773_371	R	R	R								
7	1421929_3	R	R									
8	1448748_3	S	s	S								
9	1455273_3	R	R	R								
9	1455275_3	R	R	R								
1	1427320_3	R	R	R	R			R	R	R		
2	1427186_3	R	R	S	S			S	S	s		
3	1448825_3	S	s	S								
4	1447514_3	R	R	S								
5	1408948_4	S	R	S	S			S	S	S	S	
6	1421939_3	R	R									
7	1447458_3	R	s	S								
8	1773_369	s	s	S								
9	1773_368	s	s	S								
9	1773_365	R	R	R	R			S				

Illustration of multi-layered analysis workflow



Known AMR genes uncovered by machine learning

Antibiotics

Known AMR genes

Isoniazid	$katG^{43}$, $inhA^{a20}$, $fabG1^{44}$
Rifampicin	rpoB ⁴⁵ , rpoC ^{a46} Rv3239c ⁴⁷
Ethambutol	embB ⁴⁸ , embC ¹⁷ , ubiA ^{a6} , embR ^{a17}
Pyrazinamide	pncA ⁴⁹
Streptomycin	rpsL ⁵⁰ , gidB ⁵¹
Ofloxacin	gyrA ⁵²
4-Aminosalicylic acid	folC ^{a7} , thyA ^{a53}
Ethionamide	ethA ⁵⁴ , inhA ^{a20}
Known AMR genes	$dprE1^{55}$, ald^{5} , alr^{56} , $murA^{57}$, $pks2^{58}$, $pks12^{59}$,
associated with other	$ppsA^{60}$, $ppsD^{60}$, $drrB^{61}$, $drrC^{61}$, $moeW^{55}$,
antibiotics	Rv0687 ⁶² , mshD ⁶³ , gyrB ⁵² , Rv1877 ⁶⁴ ,
	Rv0194 ⁶⁵

Gene	Drug
Rv3848	EMB, XDR
embR	EMB
Rv3129	EMB
proC	EMB
kdpC	EMB
oxcA	INH
chp2	ETA
lipD	ETA
Rv3471c	ETA, XDR, SM
mmpL11	PAS
Rv0044c	PAS
Rv0954	PAS
Rv2560	PZA
Rv2090	RIF, INH
lpqZ	RIF
Rv1597	RIF, MDR, INH
Rv1543	RIF, MDR
nuoL	MDR, PAS
dnaA	SM
yajC	SM
accD5	OFX, MDR
Rv3041c	RIF, OFX, SM, MDR
VapC21	XDR

Newly proposed AMR genes

README.md

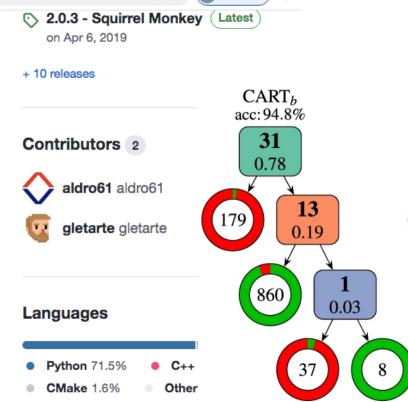


DOI 10.5281/zenodo.2630879 build passing

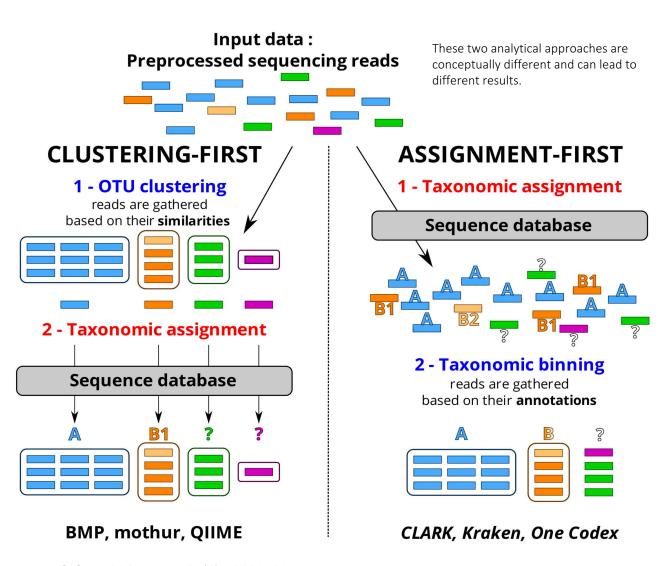
Kover is an *out-of-core* implementation of rule-based machine learning algorithms that has been tailored for genomic biomarker discovery. It produces highly interpretable models, based on k-mers, that explicitly highlight genotype-to-phenotype associations.

Introduction

Understanding the relationship between the genome of a cell and its phenotype is a central problem in precision medicine. Nonetheless, genotype-to-phenotype prediction comes with great challenges for machine learning algorithms that limit their use in this setting. The high dimensionality of the data tends to hinder generalization and challenges the scalability of most learning algorithms. Additionally, most algorithms produce models that are complex and difficult to interpret. We alleviate these limitations by proposing strong performance guarantees, based on sample compression theory, for rule-based learning algorithms that produce highly interpretable models. We show that these guarantees can be leveraged to accelerate learning and improve model interpretability. Our approach



Identification of strains from metagenome sequencing



mothur does not include taxonomic annotations at the species level.

For the QIIME integrated databases, less than 7% of Greengenes sequences and

less than 45% of SILVA sequences are annotated at the species level.

KrakenUniq - NCBI Taxonomy IDs - hundreds of strains share the same taxonomy ID



~83K complete microbial genomes, including 50K distinct bacterial genomes, 27K viral genomes, and hundreds of archaeal and eukaryotic genomes.

PLoS One. 2017 Jan 4;12(1):e0169563

 Clinical metagenomics (CMg) refers to the concept of sequencing the DNA of a clinical sample (without any prior culturing step) with the purpose of recovering clinical information

 "Salvage microbiology" is a term used for the application of molecular diagnostic techniques in the detection of bacterial DNA directly from clinical specimens

Enable the detection of pathogens and of some antibiotic resistance genes (ARGs)

Knowledge of the full drug susceptibility profile would enable tailored treatment to improve efficacy and reduce exposure to ineffective toxic drugs





☐ Antimicrobial_Resistance · Updated Apr 29, 2019 by Andy Davies

WIMP methods

Useful things to know:

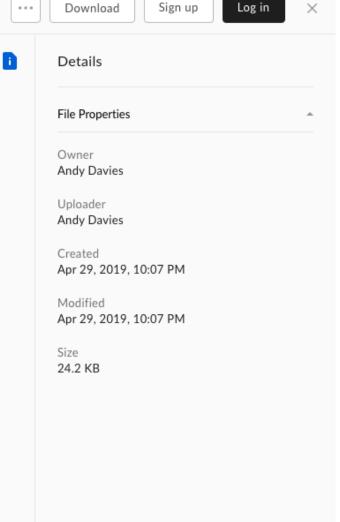
- Reads can only be assigned to organisms that are included in the database and taxonomic subtree e.g. reads will not be assigned as dog (Canis canis familiaris) as this is not included in the database however, many dog reads will likely to classified as "mammalia" as they will have homology to the human reference which is included in the database.
- The WIMP application is optimised for Oxford Nanopore Technologies DNA sequence reads (both length and error profile)
- · The application works best with genomic DNA sequences from whole genome or metagenome library preparation. Amplicon sequences such as the 16S rRNA sequence do not contain sufficient taxonomic context for this WIMP method to perform accurate sequence classification.
- The normalised gscore presented in the results is calibrated against the Zymo microbial mock community collection.

Generating WIMP database:

References / data source

WIMP classifies DNA sequence reads against a pre-generated centrifuge index. This index has been prepared from a DNA sequence collection that includes both the RefSeq microbial genome reference sequences and the primary chromosome assemblies from the human genome 1. The reference genome sequences are downloaded from the NCBI primary sequence database. Dustmasker is used to filter sequence regions of low complexity, in addition to this we mask adapter sequences. The adapters masked are provided here. The WIMP database is generated using the centrifuge-build script and is run on an amazon x1e.2xlarge instance. The centrifuge-inspect m ethod is used to collate the information for the construction of the taxonomic subtree for the genomic sequences included in the centrifuge database.

Database details:



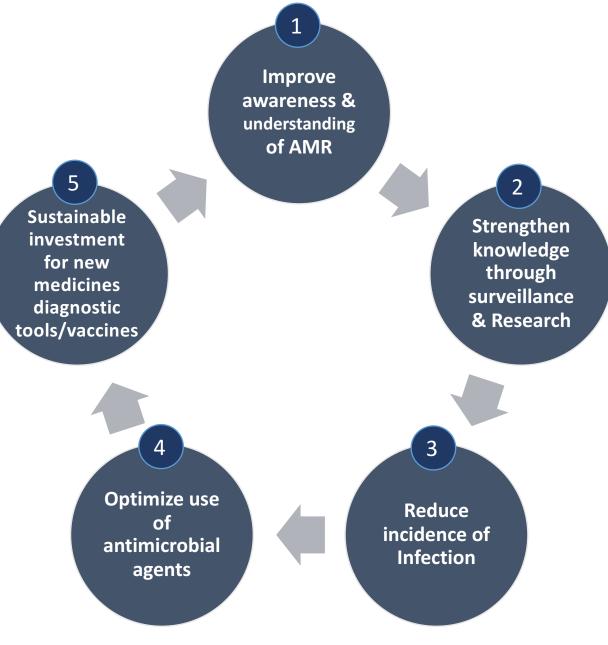
DNA sequencing as the new microscope

- It has been about 400 years since the invention of light microscopy, a technology which continues to be used and to evolve.
- It has been only 40 years since the invention of DNA sequencing; the technologies and likely to also continue to develop in the coming decades and centuries.
- It has transformed biomedical research, and is beginning to transform clinical medicine
- DNA sequencing will have a longevity and impact on par with or exceeding that of the microscope.





Sixty-eight World Health Assembly in May 2015 endorsed a global action plan to tackle antimicrobial resistance



Open Antimicrobial Resistance Awareness Drive

Abot – knowledge repository for education & awareness on AMR



A Games on education & awareness on AMR





A Chatbot will guide the Players on the most viable treatment options

Basic – Treat drug susceptible cases Advanced – Treat drug resistance cases Faster cure is the winning bet

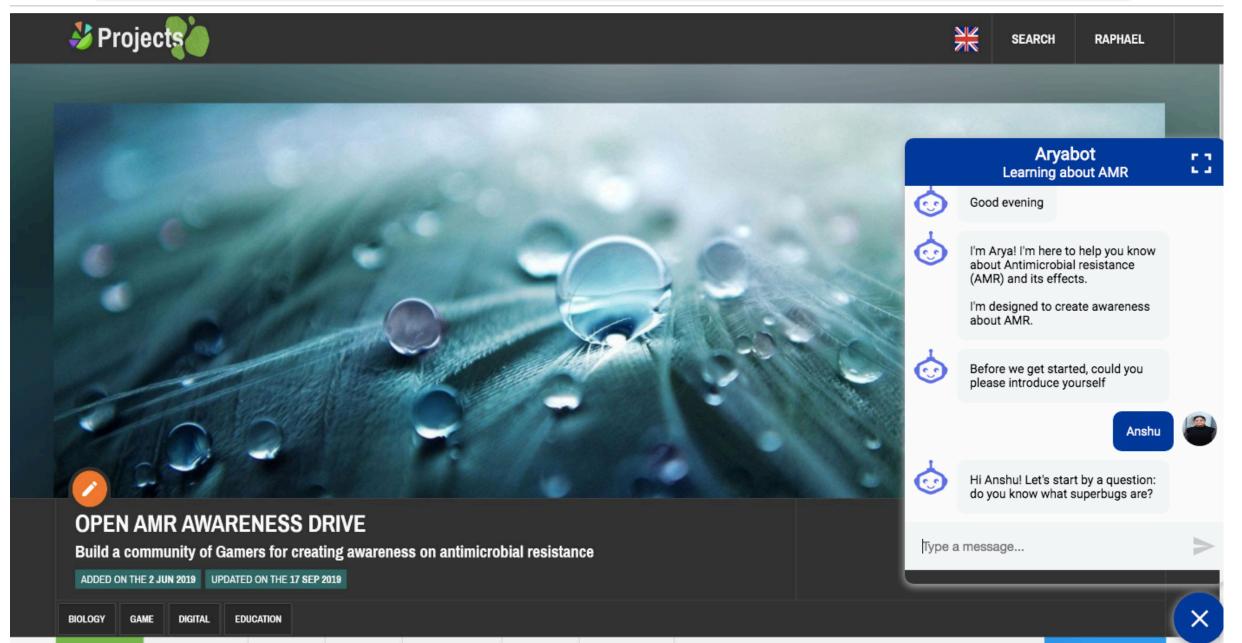


- Bot learning
- Structured knowledgebase
- Outreach

- Game Designer / creative skills
- Gamers
- Coders

https://freshdesk.com/help-desk-software/chatbot-conversations-dos-donts-blog/



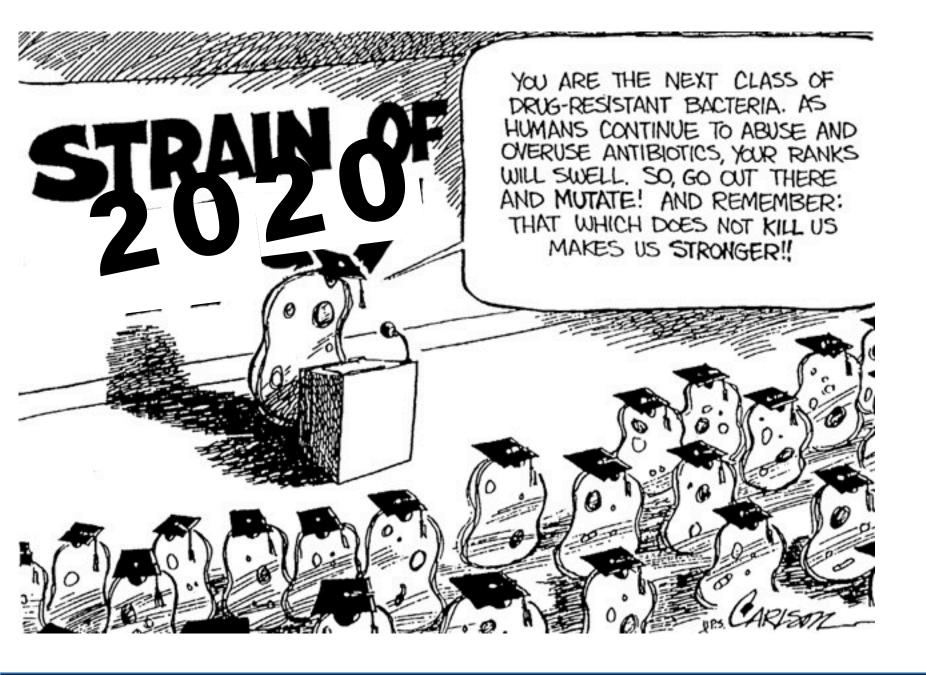






Playtest and Feedback Sessions with Aryabot





How do we convey the same in a video game?

https://slideplayer.com/slide/4685976/







Antibiotic Categorization

To address these issues, WHO developed a framework based on three different categories – Access, WAtch and Reserve – which all together forms the AWaRe categorization of antibiotics.

ACCESS GROUP

- first or second choice antibiotics
- offer the best therapeutic value, while minimizing the potential for resistance

WATCH GROUP

- first or second choice antibiotics
- only indicated for specific, limited number of infective syndromes
- more prone to be a target of antibiotic resistance and thus prioritized as targets of stewardship programs and monitoring

RESERVE GROUP

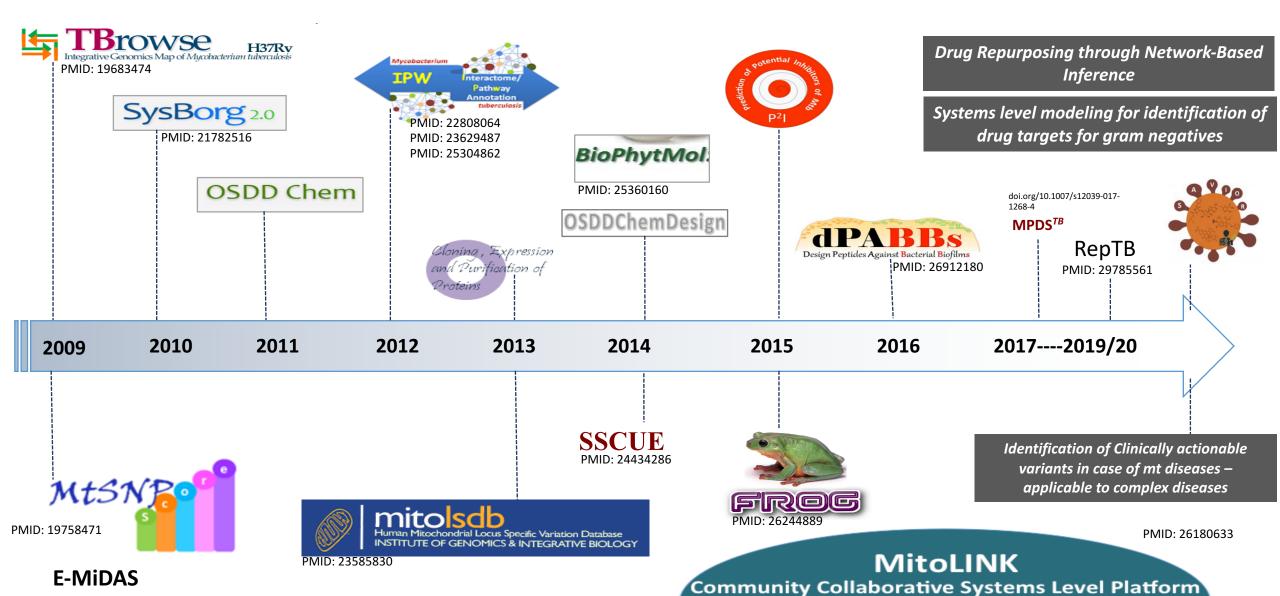
- · "last resort"
- highly selected patients (lifethreatening infections due to multi-drug resistant bacteria)
- closely monitored and prioritized as targets of stewardship programs to ensure their continued effectiveness

Game specifics are based on AWaRe

Initial game levels cannot unlock antibiotics in Watch / Reserve categories

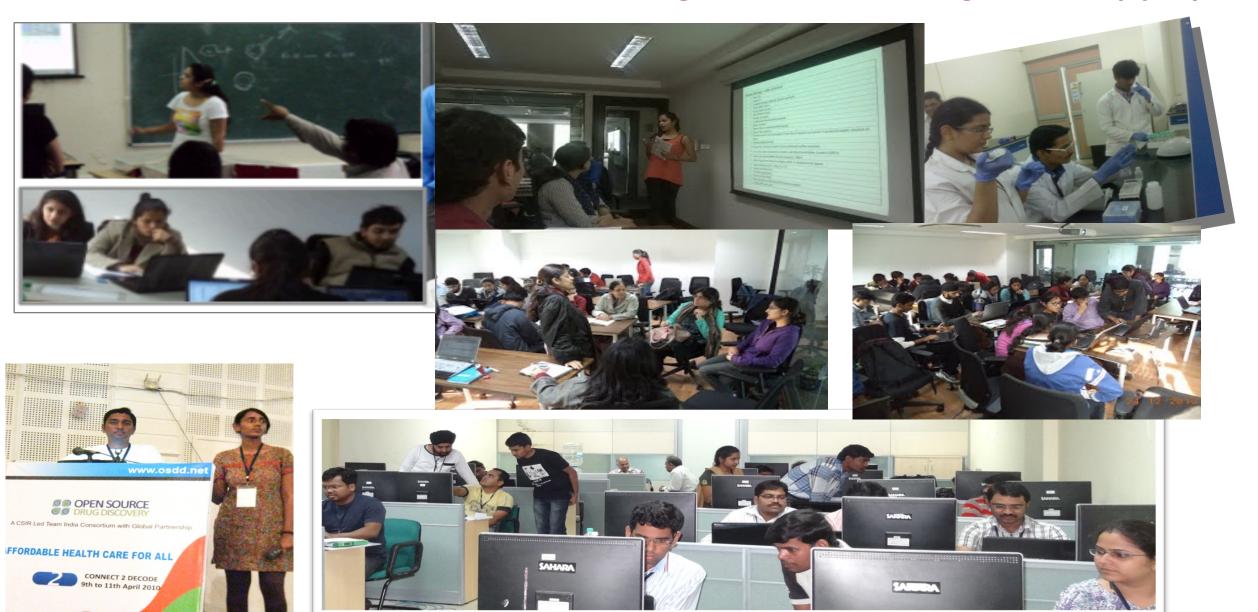
In order to unlock these, the players should accumulate points in earlier layers by assigning right antibiotic for the pathogens

Community Collaborative Projects – Crowdsourcing @ work



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Design Peptides Against Bacterial Biofilms dPABBs: A Novel in silico Approach for Predicting and Designing Anti-biofilm Peptides

SCIENTIFIC REPORTS Cambridge

OPEN dPABBs: A Novel in silico Approach for Predicting and Designing Anti-biofilm Peptides

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Arun Sharma^{1,2,*}, Pooja Gupta^{1,3,*}, Rakesh Kumar^{1,2} & Anshu Bhardwaj^{1,2} Increasingly, biofilms are being recognised for their causative role in persistent infections (like cy fibrosis, otitis media, diabetic foot ulcers) and possessmall diseases (biofilm, infected vascular cal Increasingly, biofilms are being recognised for their causative role in persistent infections (like cy fibrosis, otitis media, diabetic foot ulcers) and nosocomial diseases (biofilm-infected vascular car fibrosis, otitis media, diabetic foot ulcers) and nosocomial diseases (biofilms and their recalcitrance to conditional relevance of biofilms and their recalcitrance to conditional relevance of the conditional relevance of their recalcitrance of the conditional relevance of the conditional relevance of their recalcitrance of the conditional relevance of the fibrosis, otitis media, diabetic foot ulcers) and nosocomial diseases (biofilm-infected vascular cal implants and prosthetics). Given the clinical relevance of biofilms and their recalcitrance to convantible and prosthetics). Given the clinical relevance of biofilms and their recalcitrance to convant the same proactively sought. We have developed antibiotics, it is imperative that alternative therapeutics are proactively sought. implants and prosthetics). Given the clinical relevance of biofilms and their recalcitrance to conv antibiotics, it is imperative that alternative therapeutics are proactively sought. We have develo antibiotics, it is imperative that alternative therapeutics are proactively sought. We have develong the development of the same development of the sa dPABBs, a web server that facilitates the prediction and design of anti-biofilm peptides. The six and Weka models implemented on dPABBs were observed to identify anti-biofilm peptides on and Weka models implemented on dPABBs were observed to identify and the notifical preference of their whole amino acid composition. selected residue features and the notifical preference. and Weka models implemented on dPABBs were observed to identify anti-biofilm peptides on a constant of their whole amino acid composition, selected residue features and the positional preference of their whole amino acid composition, selected residue features and MCC of 97 50% 97 50% and residues (maximum accuracy, sensitivity, snecificity and MCC of 95, 74% 97 50% 97 50% and presidues (maximum accuracy, sensitivity, snecificity and MCC of 95, 74% 97 50% 97 50% and presidues (maximum accuracy, sensitivity, snecificity and MCC of 95, 74% 97 50 of their whole amino acid composition, selected residue features and the positional preference residues (maximum accuracy, sensitivity, specificity and MCC of 95.24%, 92.50%, 97.73% and residues (maximum accuracy, sensitivity, specificity and MCC of 95.24%, 92.50%, 97.64% and MCC of 95.24%, 92.50%, 97.73% and MCC of 95.24%, 97.73% and MCC of 95.24% and MCC of 95.2 residues (maximum accuracy, sensitivity, specificity and MCC of 95.24%, 92.50%, 97.73% and respectively, on the training datasets). On the N-terminus, it was seen that either of the cation of the cation of the training datasets. On the N-terminus, it was seen that either of the cation of the anti-hinfilm nentides. Whereas it residues. R and K. is present at all five positions in case of the anti-hinfilm nentides. respectively, on the training datasets). On the N-terminus, it was seen that either of the cation residues, R and K, is present at all five positions in case of the anti-biofilm peptides, whereas is presidues, R and K, is present at all five positions in case of the first (also anionic nolar residue S is preponderant at the first (al residues, R and K, is present at all five positions in case of the anti-biofilm peptides, whereas i peptides, the uncharged polar residue S is preponderant at the first (also anionic polar residuing the uncharged polar residue S is preponderant at the first (also anionic pola peptides, the uncharged polar residue S is preponderant at the first (also anionic polar residue third and fifth positions. Positive predictions were also obtained for 29 FDA-approved peptid third and fifth positions. Positive predictions were also obtained at their noscible renurnosi and ten antimicrobial peptides in clinical development. Indicating at their noscible renurnosi. third and fifth positions. Positive predictions were also obtained for 29 FDA-approved peptid and ten antimicrobial peptides in clinical development, indicating at their possible repurposi and ten antimicrobial peptides in clinical development, indicating at their possible repurposi and ten antimicrobial peptides in clinical development, indicating at their possible on the properties of and ten antimicrobial peptides in clinical development, indicating at their possible repurposi
biofilm therapy. dPABBs is freely accessible on: http://ab-openlab.csir.res.in/abp/antibiofilm



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Miss Pooja Gupta (2018)

Scholar-elect

MPhil Biological Science at the Department of Biochemistry

Sidney Sussex College

Biography

I was born and raised in New Delhi and am currently completing my undergraduate studies in Microbiology at the University of Delhi. Growing up on a steady diet of science fiction and popular science, I was certain by the time I was 17 that I wanted to pursue a career in research, and my journey which began under Dr. Anshu Bhardwai's guidance at the Council of Scientific and Industrial Research (CSIR), India, has been quite serendipitous since. I am keenly interested in the structure-guided fragment-based method being employed in Prof. Sir Tom Blundell's lab for antimycobacterial drug discovery. Using this method, one can not only design highly selective ligands against validated targets, but perhaps also implement the ambitious multidrug and polypharmacological strategies for more efficacious drug regimens with fewer side-effects. I hope to better understand aspects of bacterial community behaviour and explore how it may be exploited for more specific antimicrobial therapy. Learning techniques in Structural Biochemistry, principles of Drug Design, Systems-level approaches and Synthetic Biology could eventually enable an integration of these fields to study interactions in microbial communities, along with emergent phenomena like biofilm formation and drug resistance. I would also like to teach students and dedicate time to science communication for public engagement, which I believe is crucial to the larger scientific cause. It would be an exciting challenge to try and capture popular imagination with advanced



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