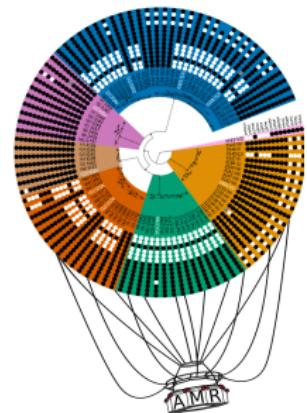


Around the resistome in 80 ways:

an empirical evaluation of antimicrobial
resistance gene detection methods



Finlay Maguire

finlaymaguire@gmail.com

December 2, 2019

Faculty of Computer Science, Dalhousie University

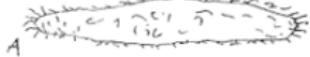
Table of contents

1. Background
2. Why do we care about AMR?
3. Targeted sequencing
4. Genomics
5. Metagenomics
6. Metagenomic-Assembled Genomes

Background

Evolution of Eukaryotic Endosymbioses

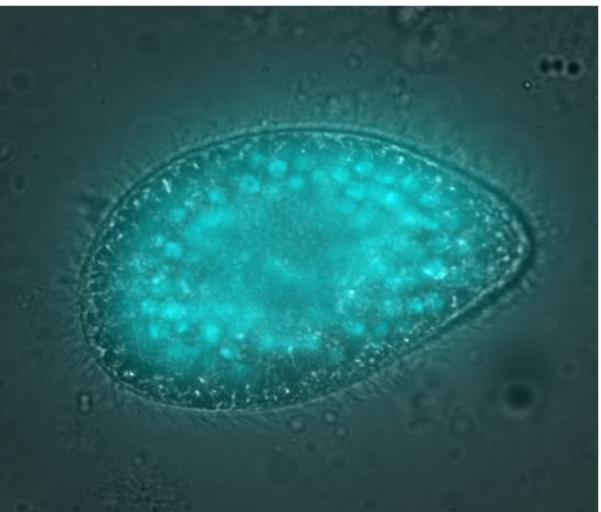
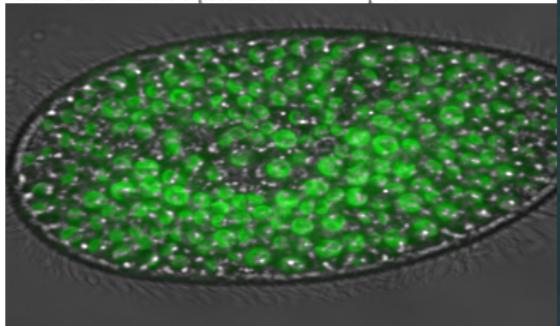
peu. Et par deux fois j'ay vu dans cette mesme eau un animal dix que ces autres qui avoit des pieds tout le long du corps, et estoit



A

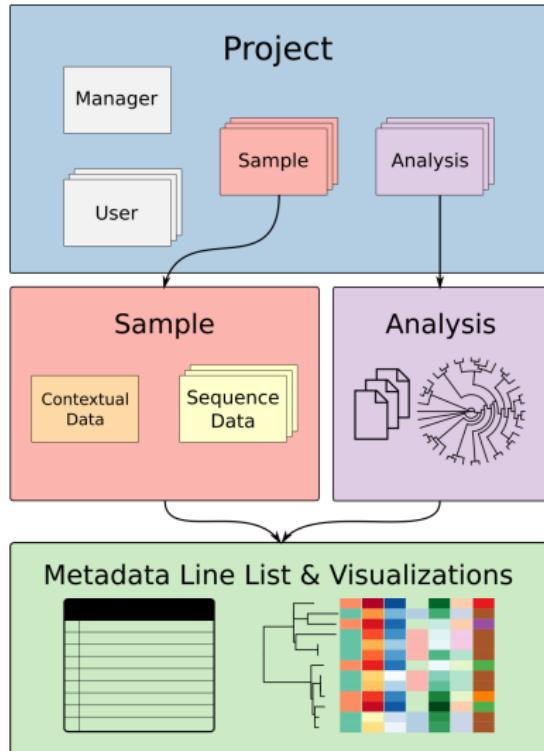
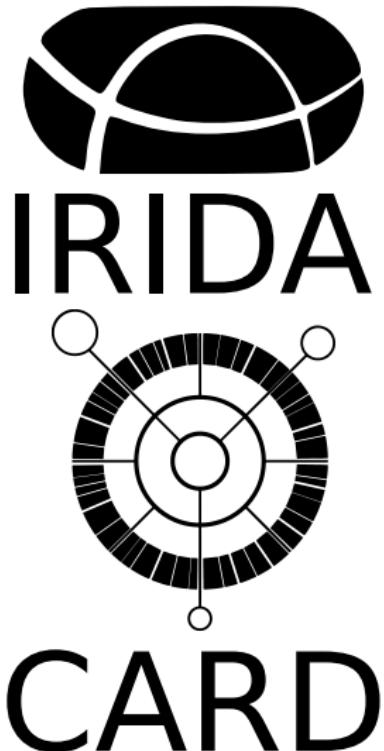
Les 4 ou 5 pieds du c
fans cessé quand mesme
en repos. Il courroit v
autres, et se tournoit et
l'eau. Hartsoecker m'a

trouvé de la mesme espece in femme corrupto.



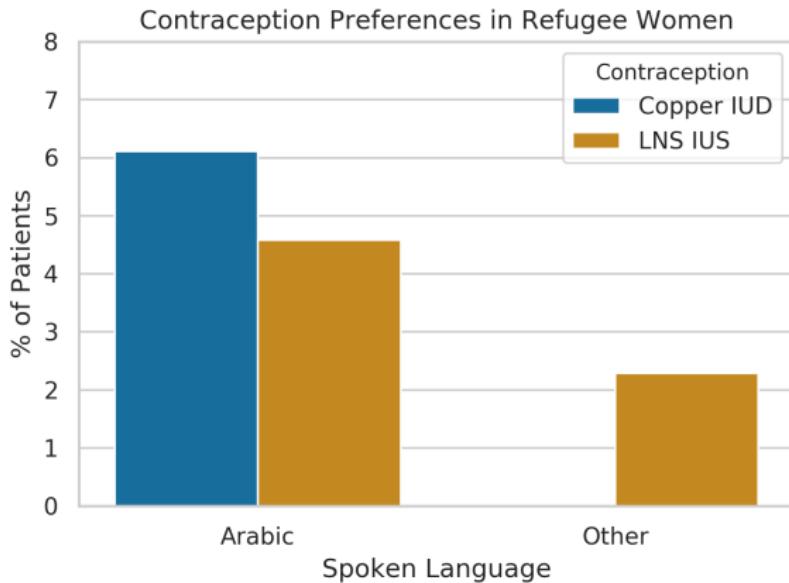
(Maguire, 2016)

Antimicrobial Resistance



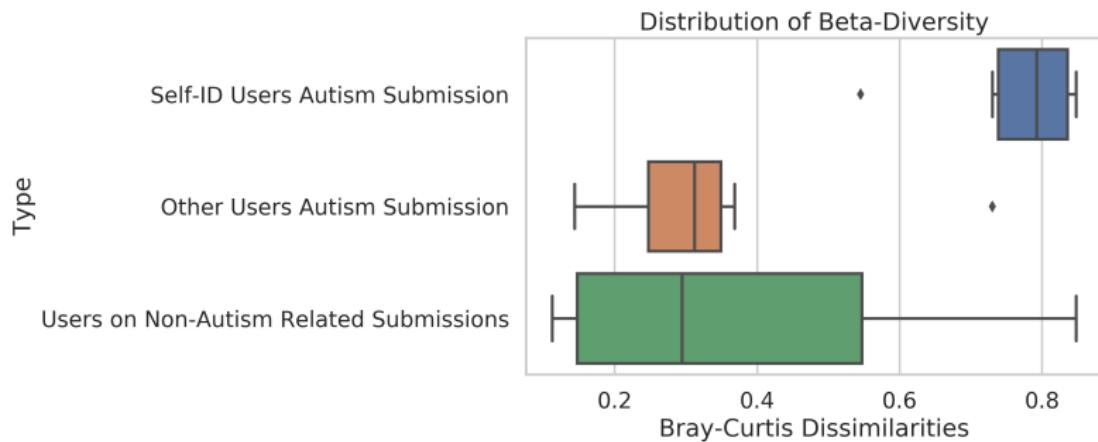
(Matthews et al., 2018)

Epidemiology



(Stairs et al., 2019)

Sociology?



Sociology?

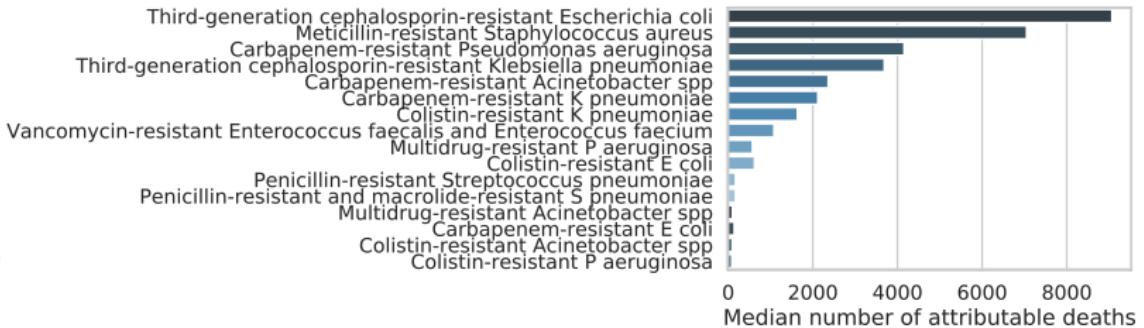
Congratulations, your application to the SSHRC Explore Grants competition has been awarded.

Project Title: NEETs, Incels, and Wizards: The Experiences of Socially Isolated Men

Why do we care about AMR?

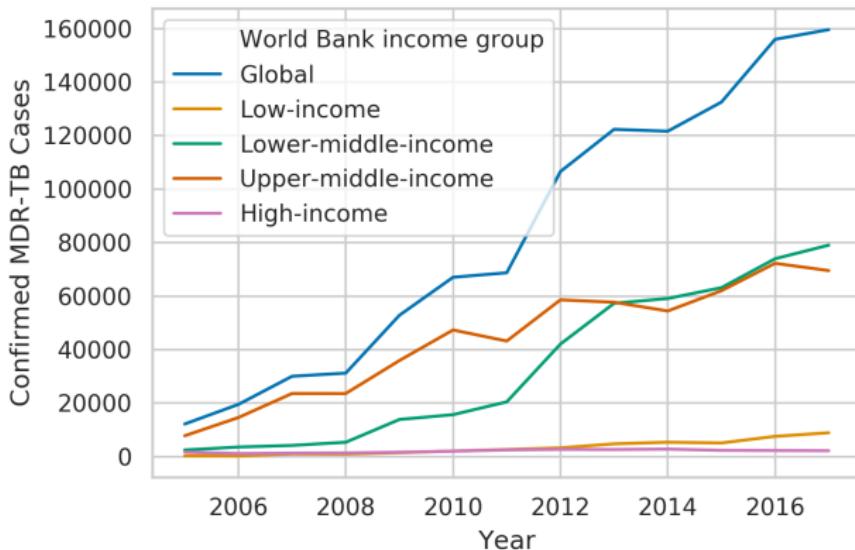
AMR is currently a problem

EU/EEA Resistant Infections



2015 EU/EEA: 33,110 deaths, Data from (Cassini et al., 2019).

AMR is growing



WHO Global Health Observatory Data Repository.

What can we do about it?

Improve surveillance

- Locally: information would help improve patient health.

Improve surveillance

- Locally: information would help improve patient health.
- Nationally: health policies and responses to emergencies.

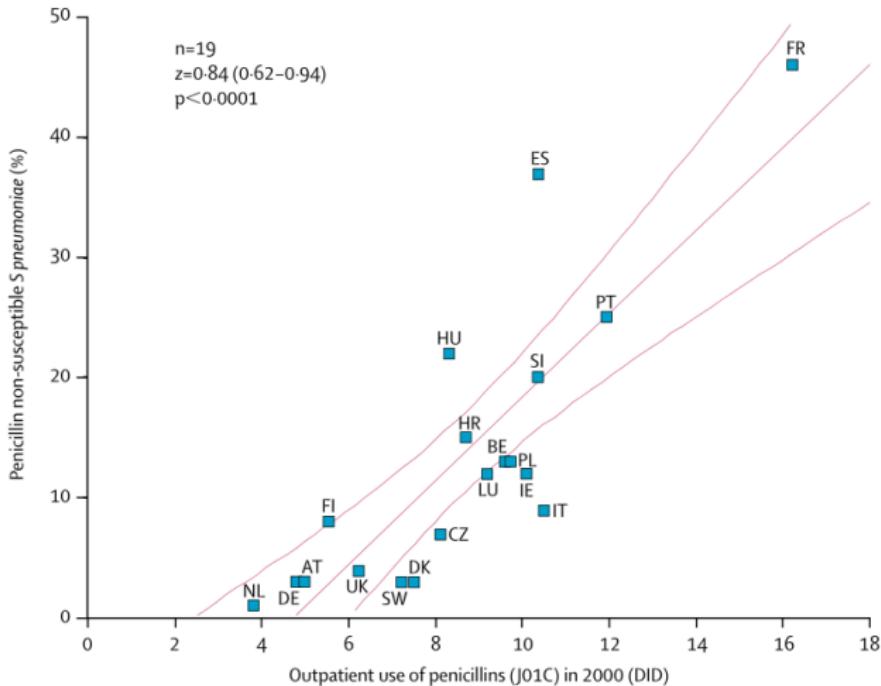
Improve surveillance

- Locally: information would help improve patient health.
- Nationally: health policies and responses to emergencies.
- Globally: emerging threats and long-term trends.

Improve surveillance

- Locally: information would help improve patient health.
- Nationally: health policies and responses to emergencies.
- Globally: emerging threats and long-term trends.
- Scientifically: better understanding of underlying biology.

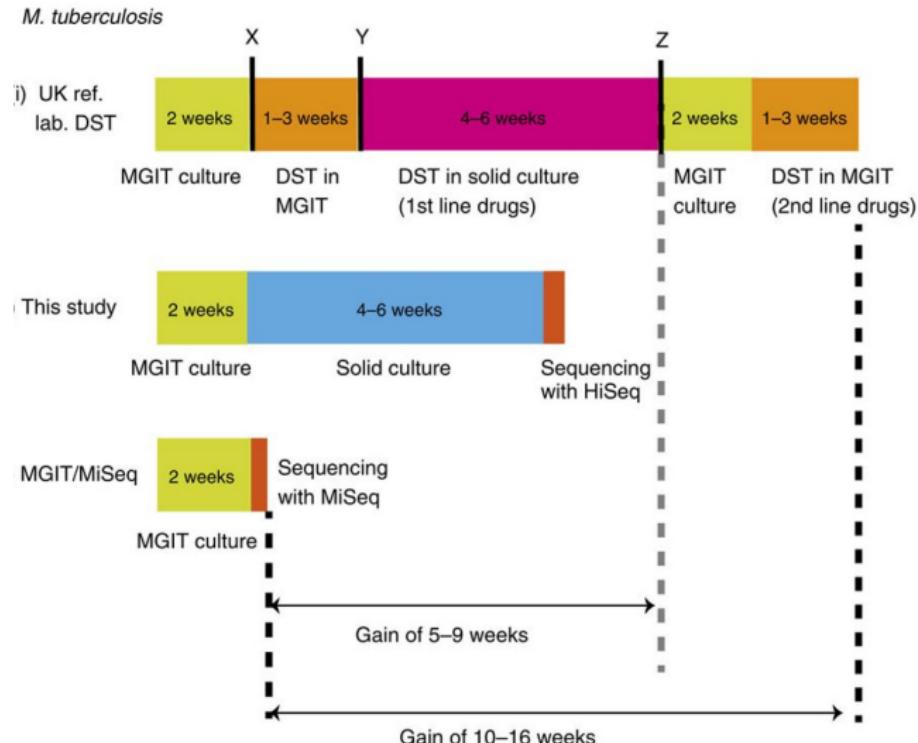
Improve diagnostics



(Goossens et al., 2005)

How do we do this?

Phenotypically?



(Bradley et al., 2015)

DNA sequencing

- DNA is relatively tractable and stable
- Sequencing technology is mature
- Represents the substrate of evolution

Downside of DNA: capacity not expression

E. coli gene regulatory networks are
inconsistent with gene expression data ⚡

Simon J Larsen ✉, Richard Röttger, Harald H H W Schmidt, Jan Baumbach

Nucleic Acids Research, Volume 47, Issue 1, 10 January 2019, Pages 85–92,

Downside of DNA: capacity not expression

Random sequences rapidly evolve into de novo promoters

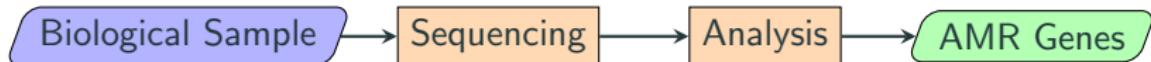
Avihu H. Yona , Eric J. Alm & Jeff Gore 

Nature Communications 9, Article number: 1530 (2018) | [Cite this article](#)

- 10% of random sequences can serve as active promoters
- 60% of random sequences can modulate expression with only one mutation

Which DNA sequencing method?

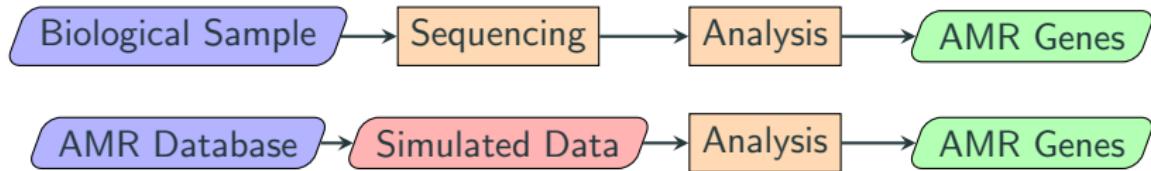
Choosing a method



Additional factors:

- Does method provide other information?
- Cost/experimental considerations

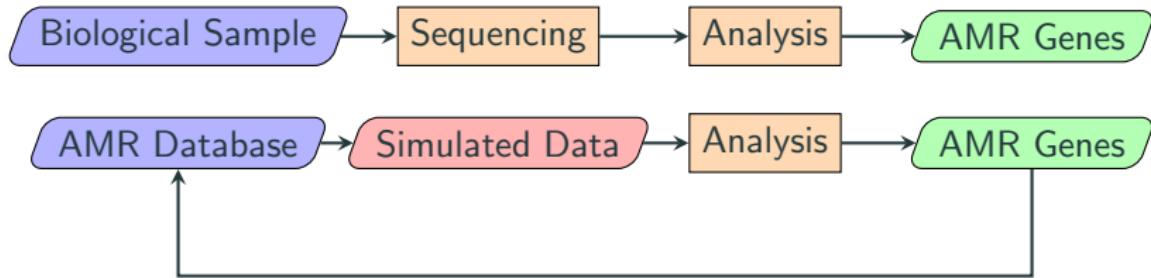
Choosing a method



Additional factors:

- Does method provide other information?
- Cost/experimental considerations

Choosing a method

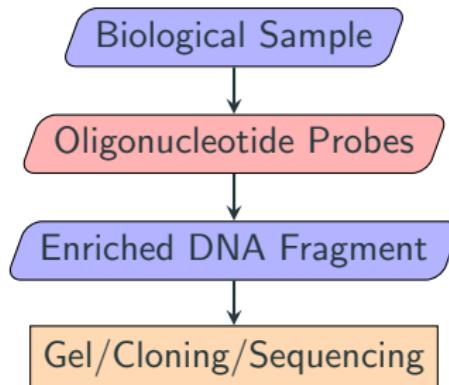


Additional factors:

- Does method provide other information?
- Cost/experimental considerations

Targeted sequencing

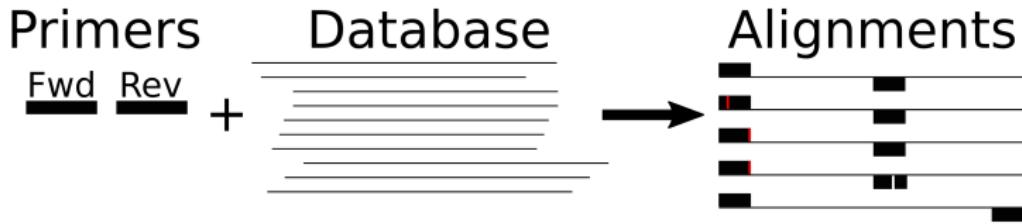
Targeted sequencing



- Cheap/simple infrastructure
- Multiple sample types
- Low input requirements

Choosing and evaluating primers

Testing primers computationally



github.com/mwhall/VAware: Mike Hall

Needleman-Wunsch alignments:

- Perfect: no mismatches, insert < 1500
- Intermediate: (1-2 minor mismatches)
- Low: (2-4 minor; 0-1 major - terminal, gaps)
- Missed: (> 4 minor; > 1 major)

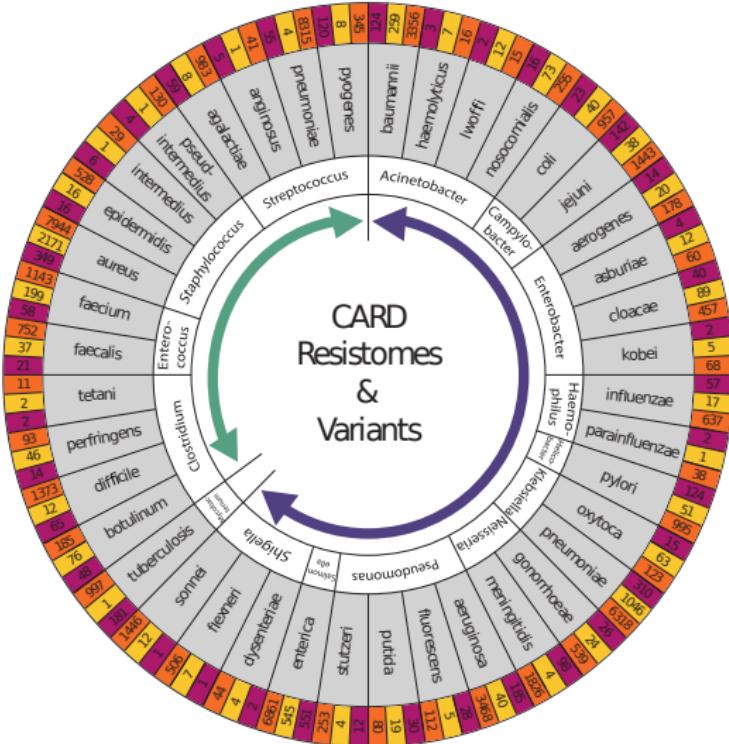
Which primers?

List of primers for detection of antimicrobial resistance genes

Antibiotic	Group	Gene/locus	Primer name	Internal number	Sequence	Temperature (°C)	Reference
Beta-lactam	TEM	All	TEM front PI	Primer 757	5'-GGGAAACCUUATTTG-3'	55	Oleow, I., H. Berent, and F. M. Aarsting. 2004. Prevalence of Beta-lactamase among ampicillin-resistant <i>Escherichia coli</i> and <i>Salmonella</i> isolated from food animals in Denmark. <i>Microbiol Res.</i> 159:136-140.
			TEM-C-R-ey	Primer 688	5'-ACC AAT GCT TAA TCA GTC AG-3'		Moedas, A. and Gualdeiros, L. Transmission of <i>In</i> N Plasmids Carrying <i>bla</i> CTX-M-1 between Commercial <i>Escherichia coli</i> in Pigs and Farm Workers. <i>Antimicrob Agents and Chemotherap.</i> 2009; 53:1709-1711.
CTX	M-EI	cIM U1	Primer 1354		5'-AATGCGAGYACCAAGTAABUTKATGCC-3'	60	Braaten, H., D. Moers, K. Veldman, I. Oleow and F. M. Aarsting. 2006. Beta-lactamase among Extended-spectrum Beta-lactamase resistant (ESBL) <i>Salmonella</i> from poultry, poultry products and human patients in The Netherlands. <i>J. Antimicrob Chemother.</i> 57:103-107.
		CTX-M-U2new	Primer 1580		5'-TGGGTRAAARTGTTGACAGAAYSAGCGG-3'		Harmsen, M., Meiners, M., Rombach, C., Becker, E.L., Drew, S.V., Kader, C., Haarmann, H., Conraths, M., Moers, D., Thiel, J.E., Angulo, F.J., Aarsting, F.M. 2009. Emergence of Multidrug Resistant <i>Salmonella</i> Concord Infections in Europe and the United States in Chickens Adapted From Ethiopia, 2003-2007. <i>Pediatr Infect Dis J.</i> 28:814-818.
	CTX-M group	ctx-M-15 front PI	Primer 1537		5'-CCATGGTTAAAAATCACTCG-3'		Moedas, A. and Gualdeiros, L. Transmission of <i>In</i> N Plasmids Carrying <i>bla</i> CTX-M-1 between Commercial <i>Escherichia coli</i> in Pigs and Farm Workers. <i>Antimicrob Agents and Chemotherap.</i> 2009; 53:1709-1711.

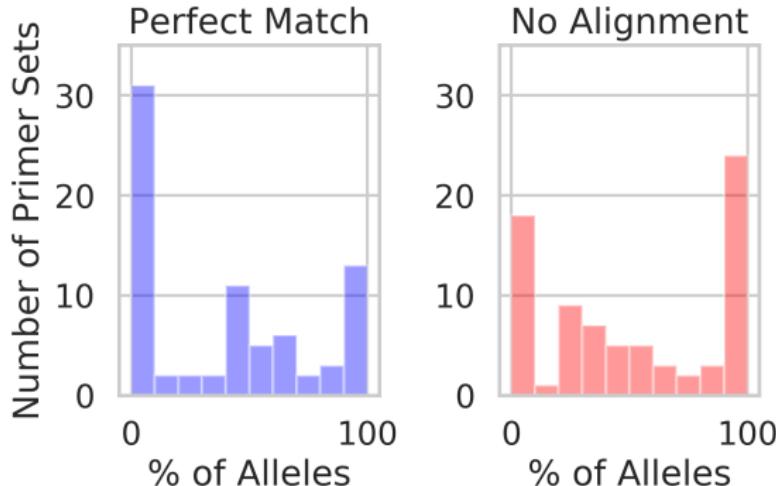
European Committee on Antimicrobial Susceptibility Testing: 78 PCR Primer Sets

Which AMR genes?



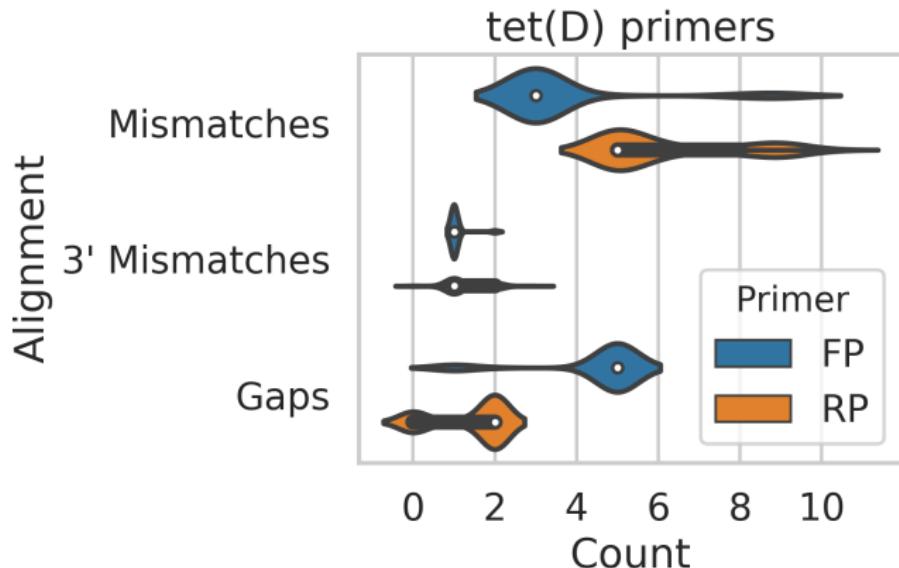
How well do these primers work?

Surprisingly poorly



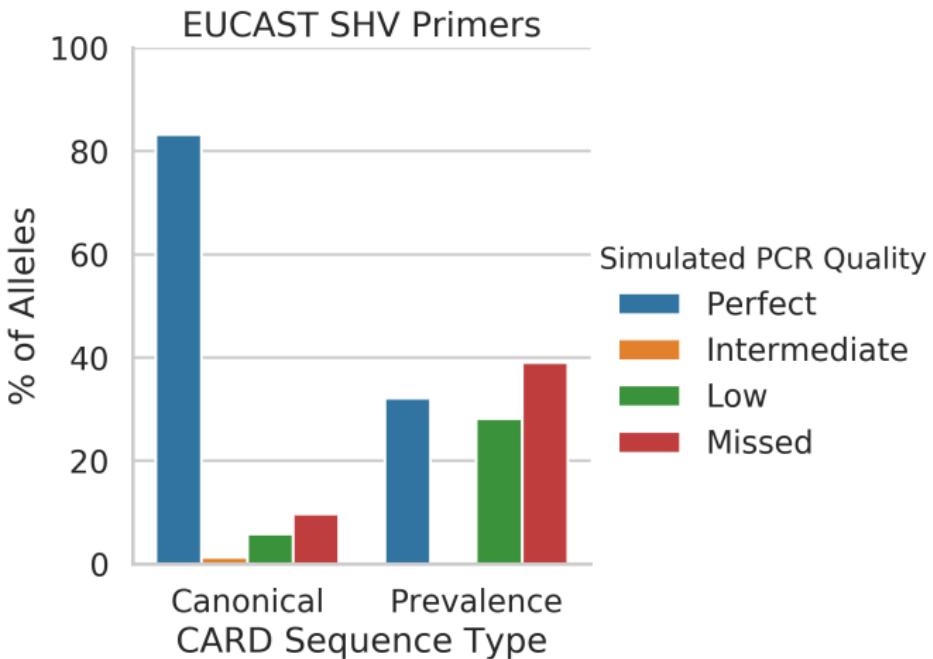
- Many aminoglycosides and tetracycline resistance genes totally missed
- Caveat: needs experimental validation

Lots of serious mismatches



No primer alignment in 27.58% of *tetD* alleles

Stagnation of primers



off-target hits (1 mismatch in RP) to *LEN-3*, *LEN-4*

Can we improve on this?

Designing probes with up-to-date AMR allele diversity



Antimicrobial Agents
and Chemotherapy

Mechanisms of Resistance

Capturing the Resistome: A targeted capture method to reveal antibiotic resistance determinants in metagenomes

Allison K. Guitor, Amogelang R. Raphenya, Jennifer Klunk, Melanie Kuch, Brian Alcock,
Michael G. Surette, Andrew G. McArthur, Hendrik N. Poinar, Gerard D. Wright

DOI: 10.1128/AAC.01324-19

(Guitor et al., 2019)

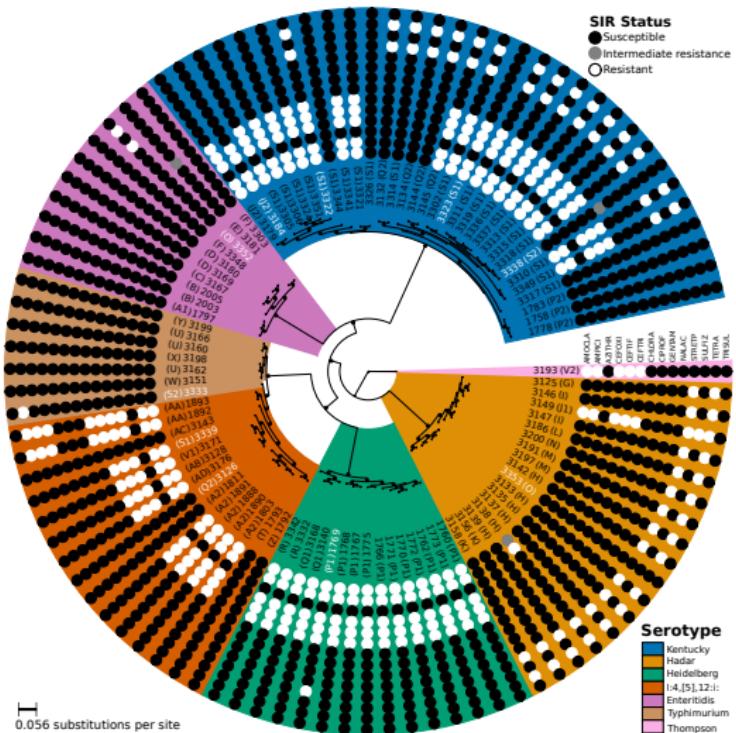
Downsides of targeted-approaches

- *a priori* target decisions
- Need constantly updated
- No easy genomic context
- No easy source-genome attribution

Why do we care about context?

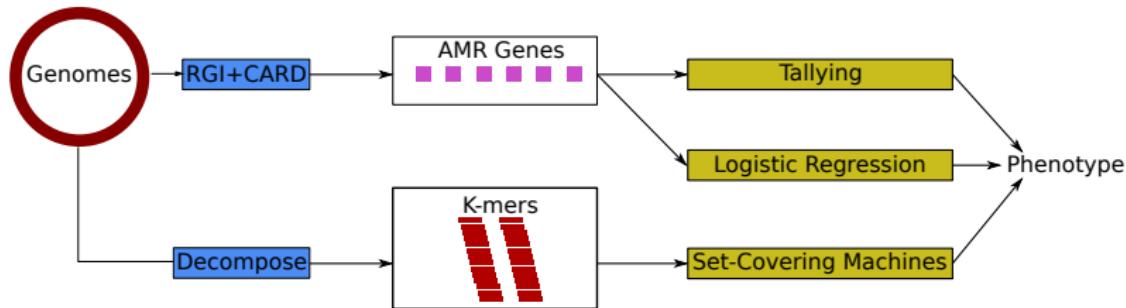
Genomics

Case-study on strengths of genomics



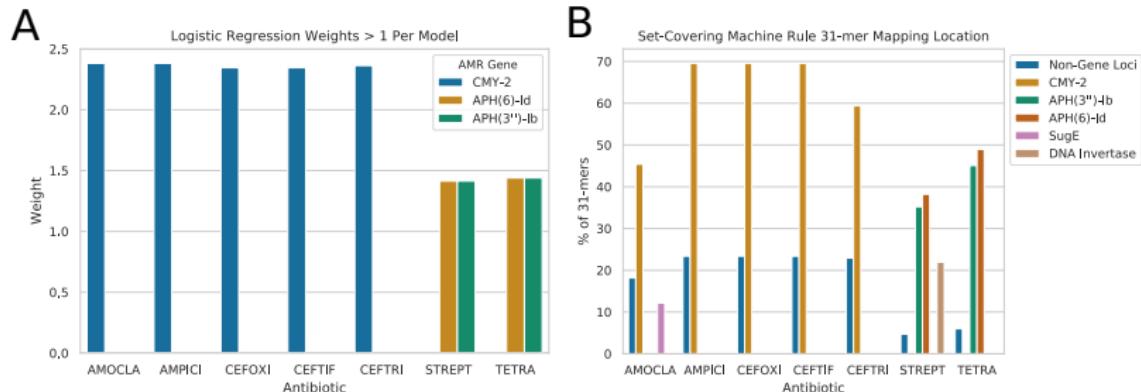
(Maguire et al., 2019)

Phenotype prediction modelling



(Maguire et al., 2019)

Genomes allow gene-free models



(Maguire et al., 2019)

Generate co-selection hypotheses

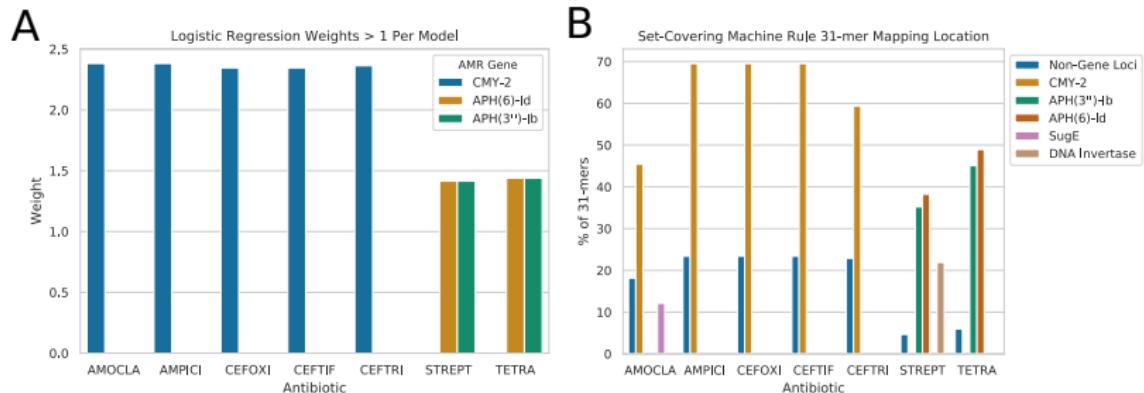
Appl Environ Microbiol. 2011 Jul;77(13):4486-93. doi: 10.1128/AEM.02788-10. Epub 2011 May 20.

Selection pressure required for long-term persistence of blaCMY-2-positive IncA/C plasmids.

Subbiah M¹, Top EM, Shah DH, Call DR.

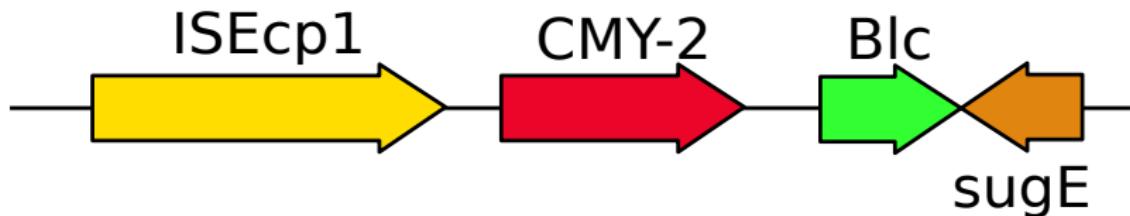
(Maguire et al., 2019)

Generate co-selection hypotheses



(Maguire et al., 2019)

Generate co-selection hypotheses



(Maguire et al., 2019)

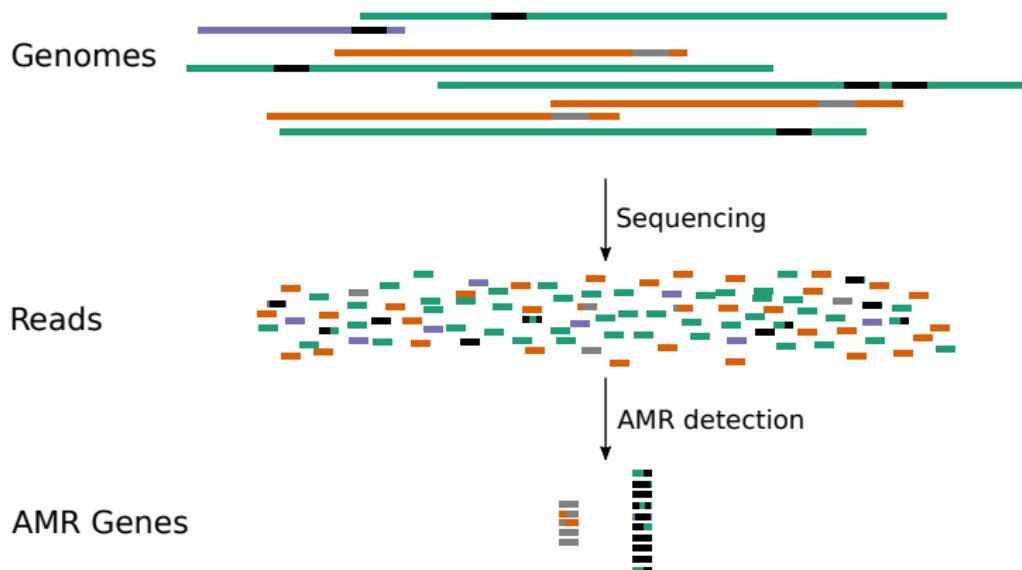
Downsides of genomics

We need genomes to identify previously unknown factors, but:

- Culturing is expensive, time-consuming, and difficult
- Single cell methods are noisy and analytically complex
- Only profile 'one' genome per sample

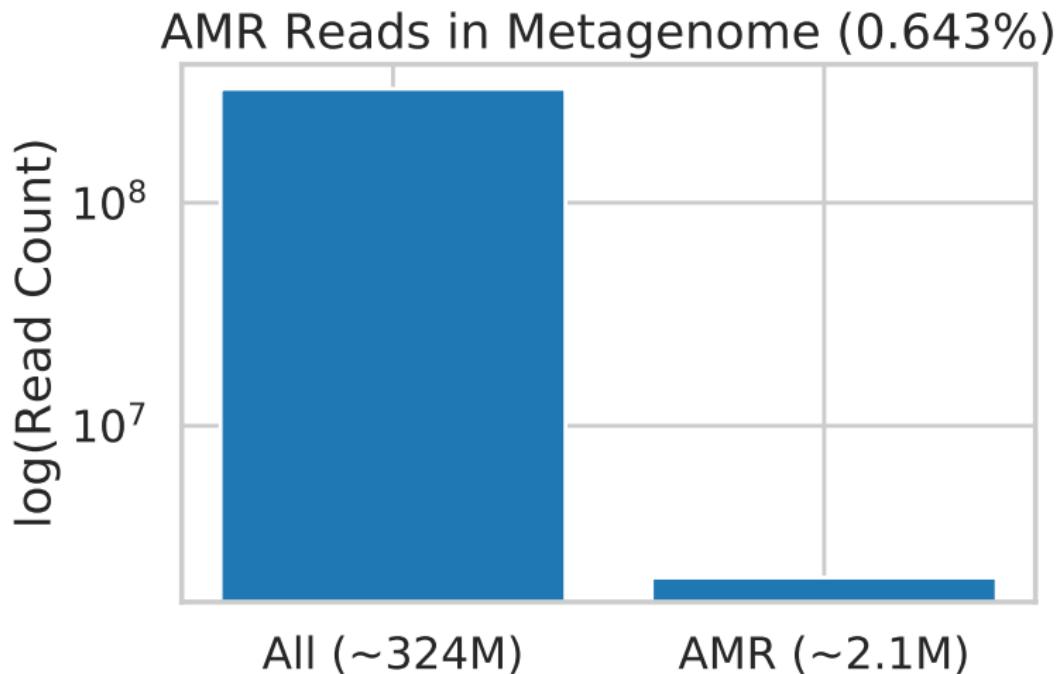
Metagenomics

Read-based AMR Metagenomics



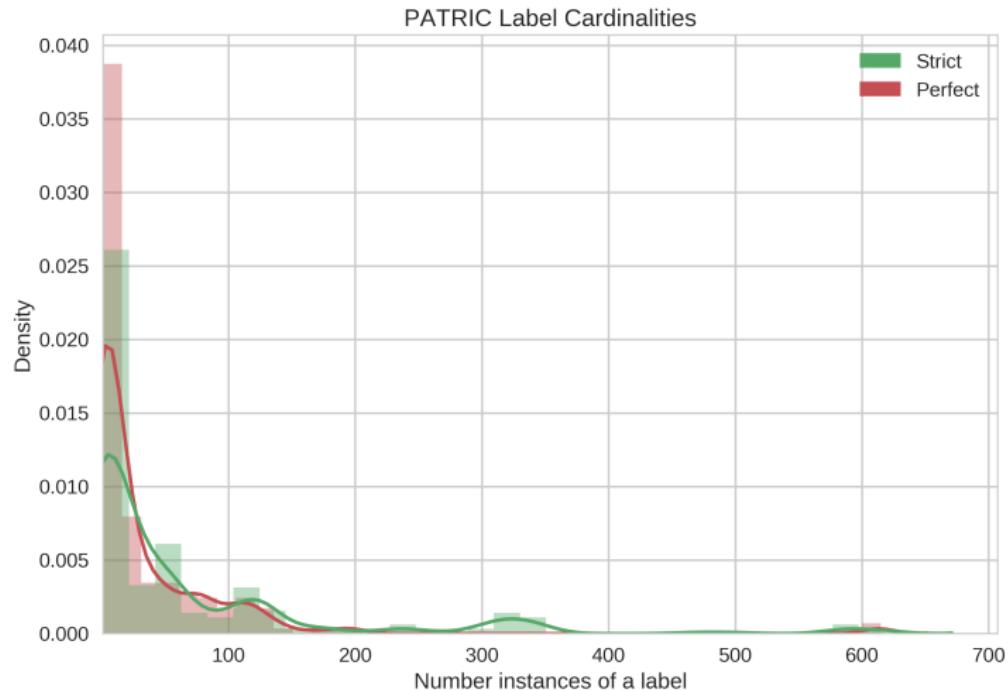
Difficulties of metagenomics

AMR genes are rare genetically



2184 CARD-prevalence genomes at 1-10X abundance

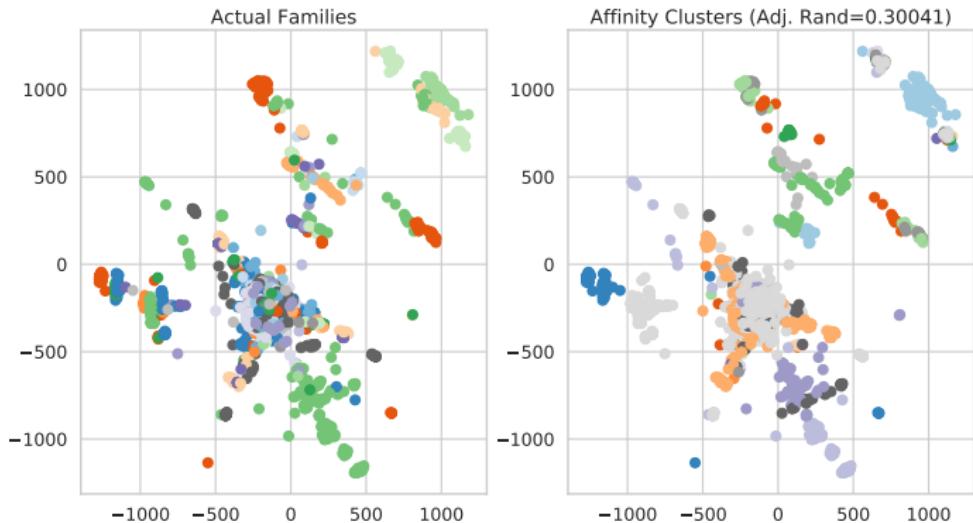
AMR genes have wildly different abundances



1236 AMR PATRIC genomes

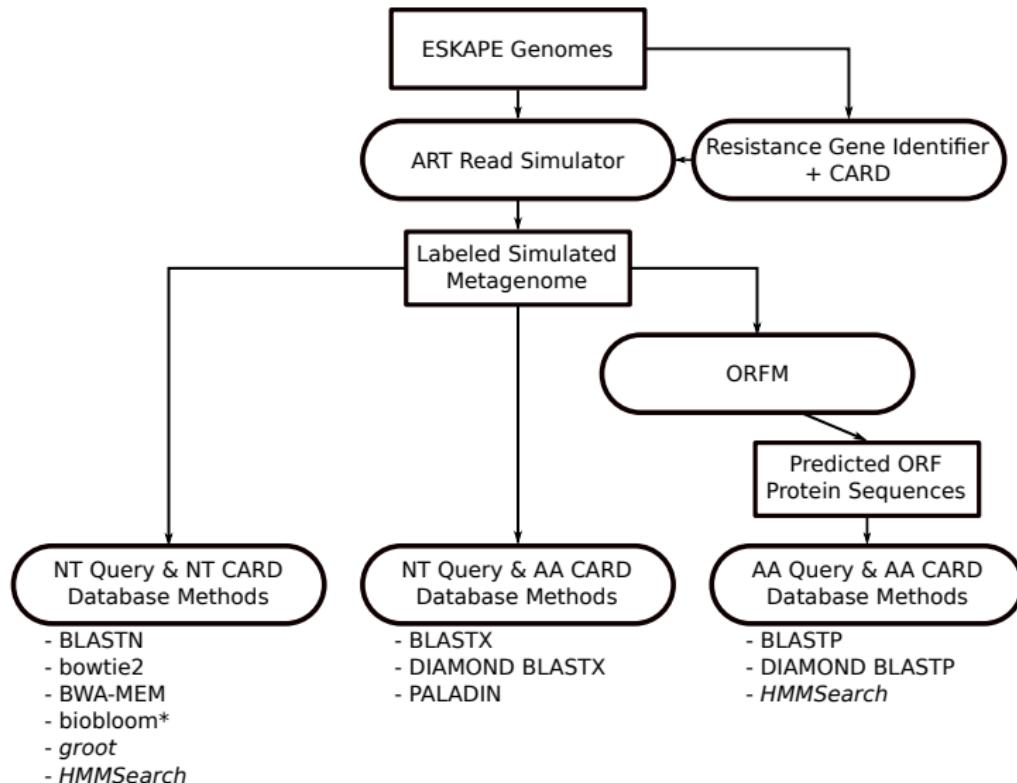
AMR sequence space overlaps

MDS of CARD Proteins BLASTP-%ID

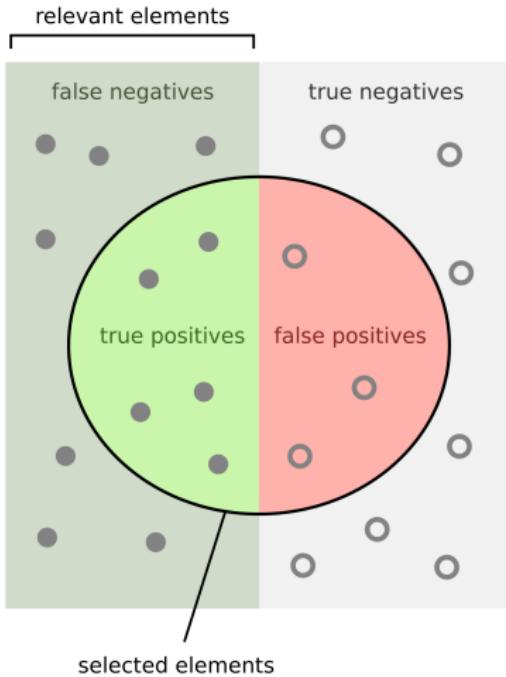


Choosing an analysis approach

Simulate data and compare tools



Terminology refresher



How many selected items are relevant?

$$\text{Precision} = \frac{\text{true positives}}{\text{selected elements}}$$

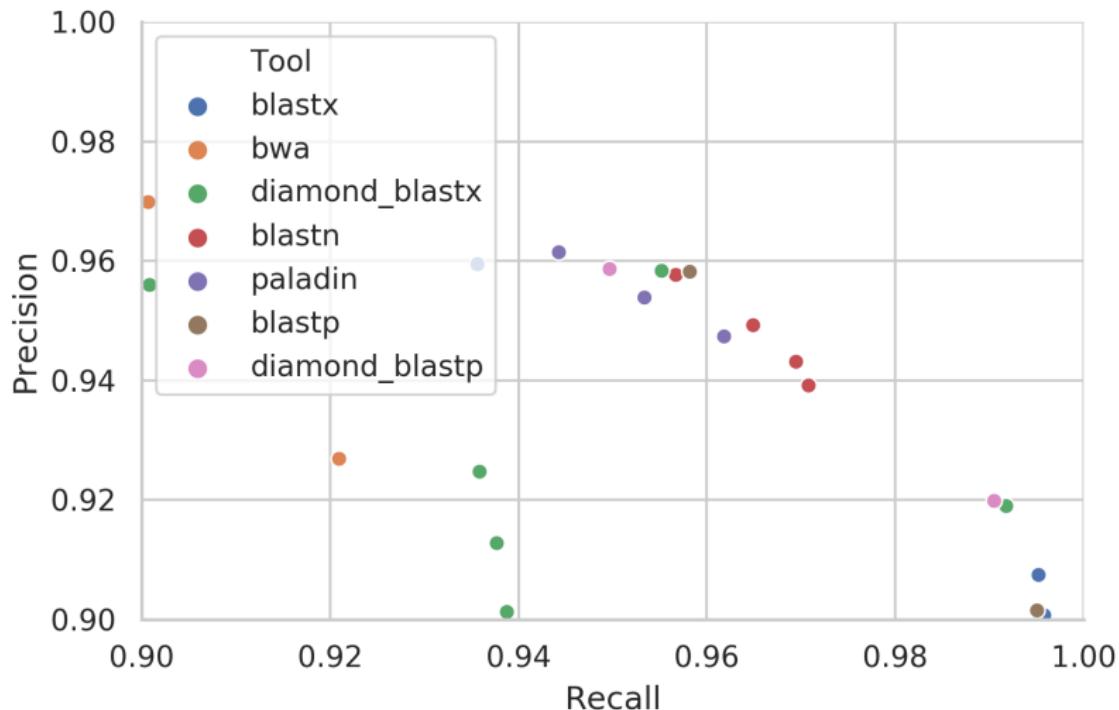
How many relevant items are selected?

$$\text{Recall} = \frac{\text{true positives}}{\text{relevant elements}}$$

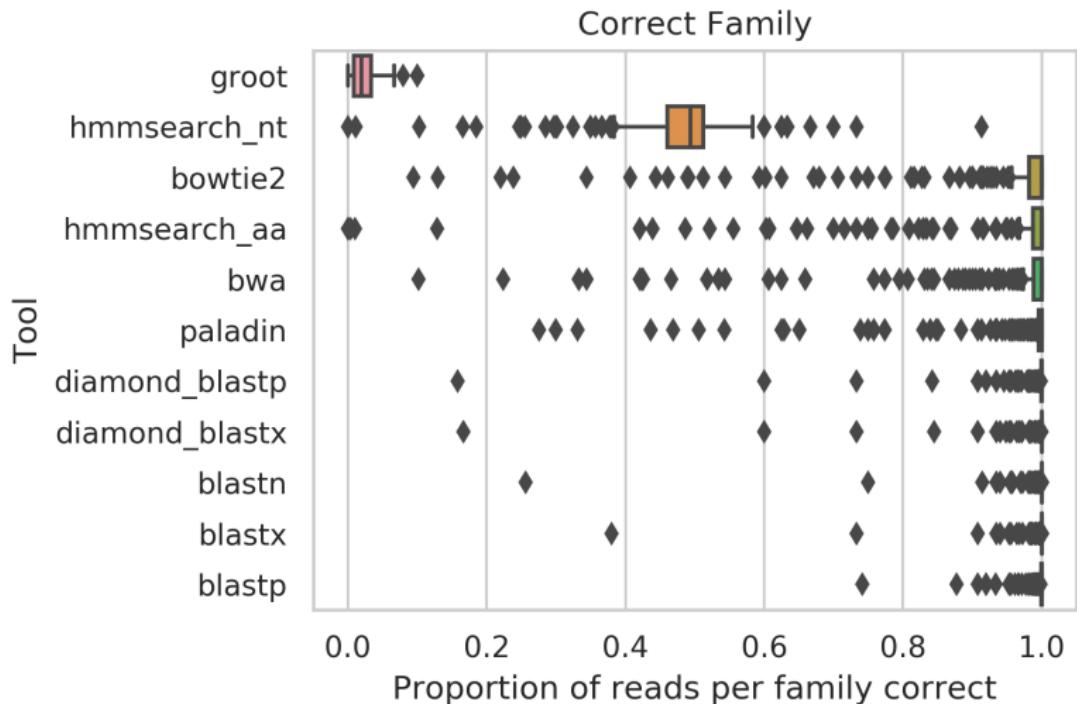
bit.ly/2pZzxJU

**How well do different methods
do?**

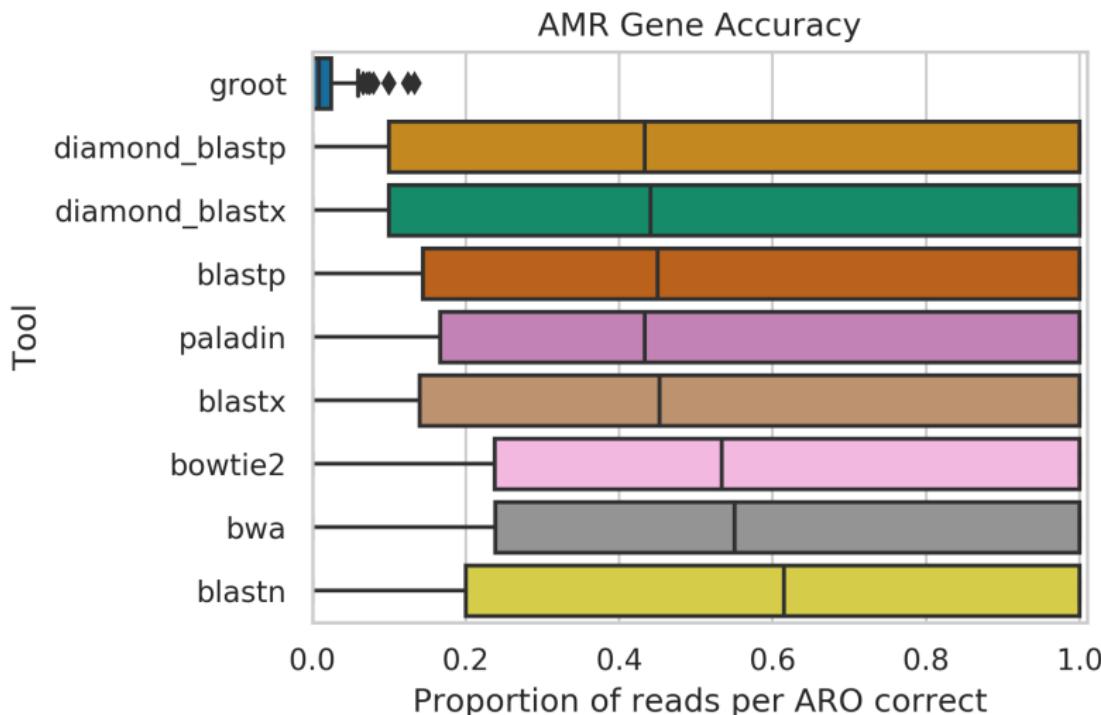
We can find reads from AMR genes



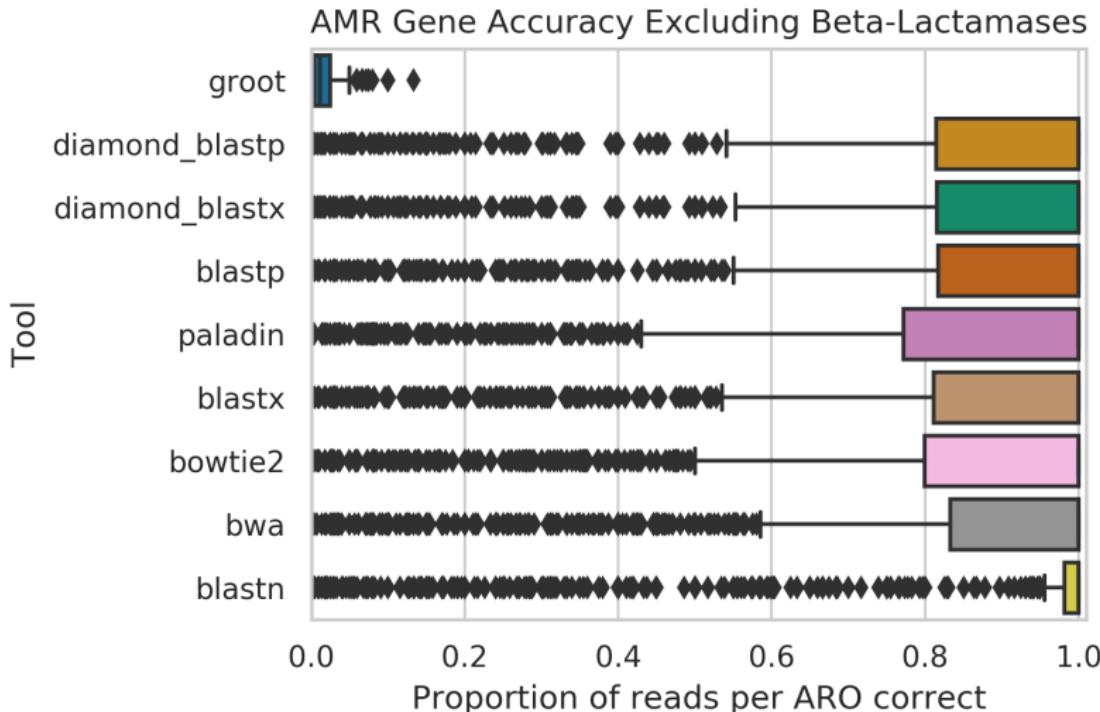
We can mostly identify which family



We cannot identify which specific gene



Highly similar families to blame

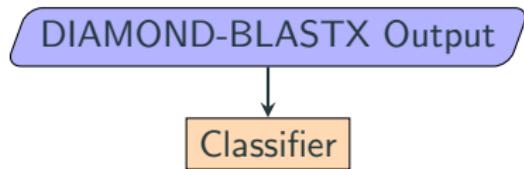


Is there any way to improve this?

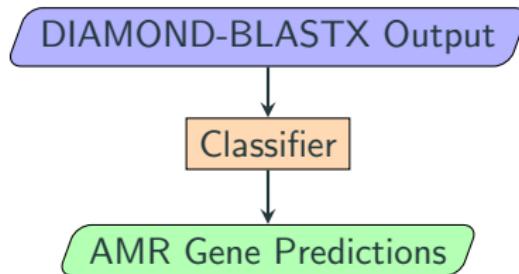
Statistical/Machine-Learning Correction

DIAMOND-BLASTX Output

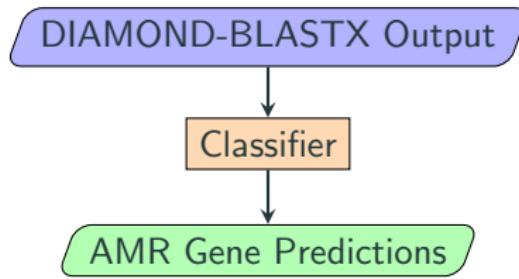
Statistical/Machine-Learning Correction



Statistical/Machine-Learning Correction

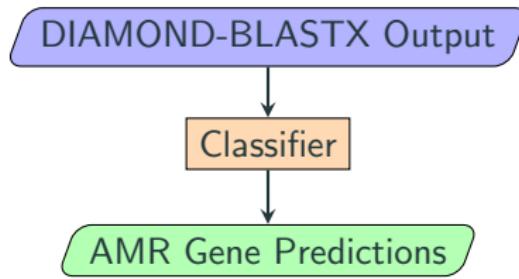


Statistical/Machine-Learning Correction



Average Precision: 0.63

Statistical/Machine-Learning Correction

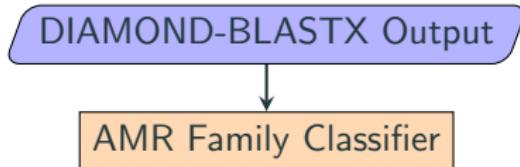


Average Precision: 0.63 %

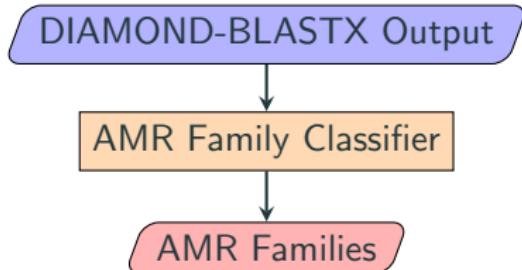
Revised classifier structure: exploiting the ARO

DIAMOND-BLASTX Output

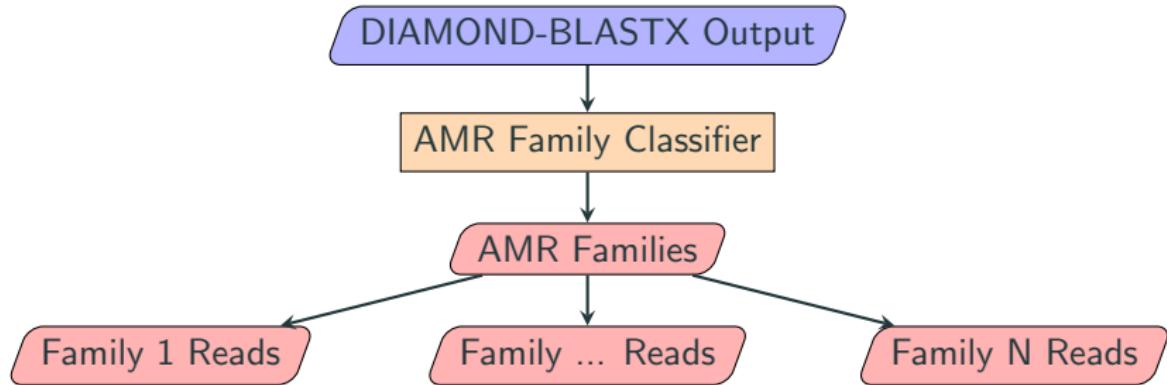
Revised classifier structure: exploiting the ARO



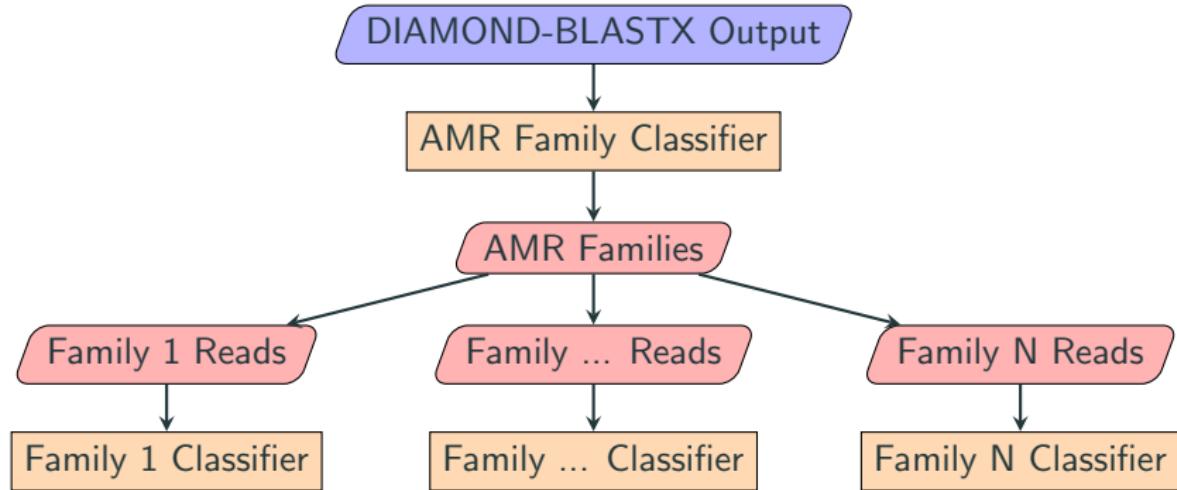
Revised classifier structure: exploiting the ARO



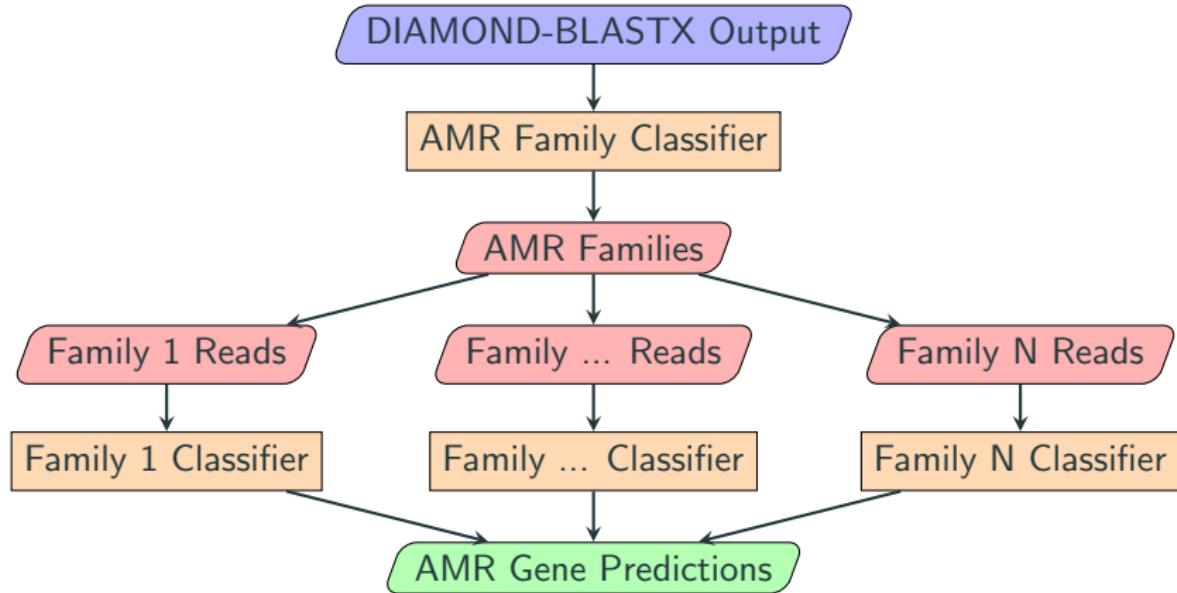
Revised classifier structure: exploiting the ARO



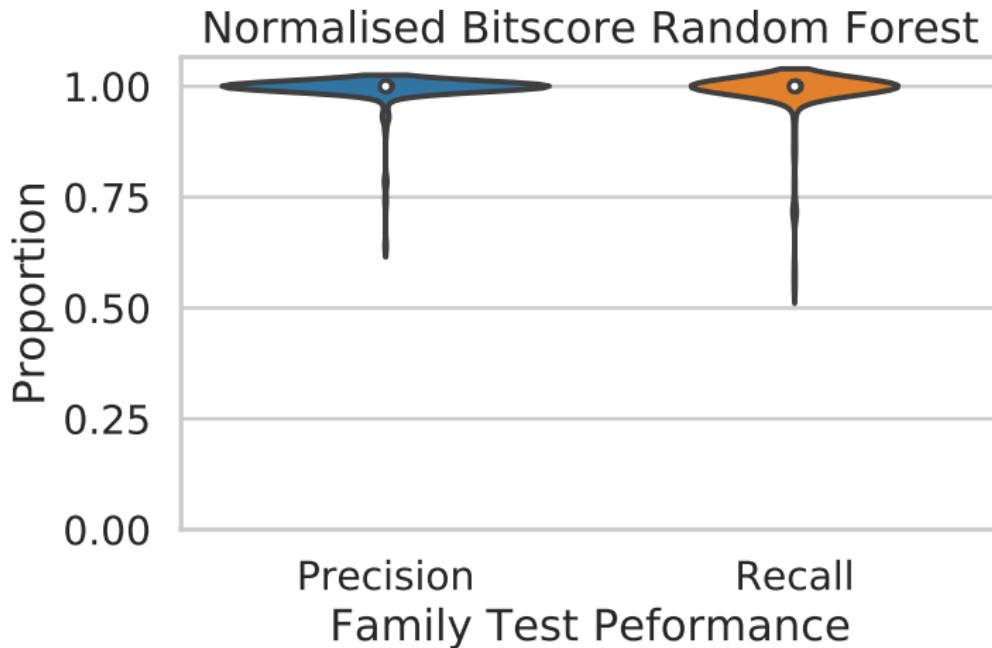
Revised classifier structure: exploiting the ARO



Revised classifier structure: exploiting the ARO

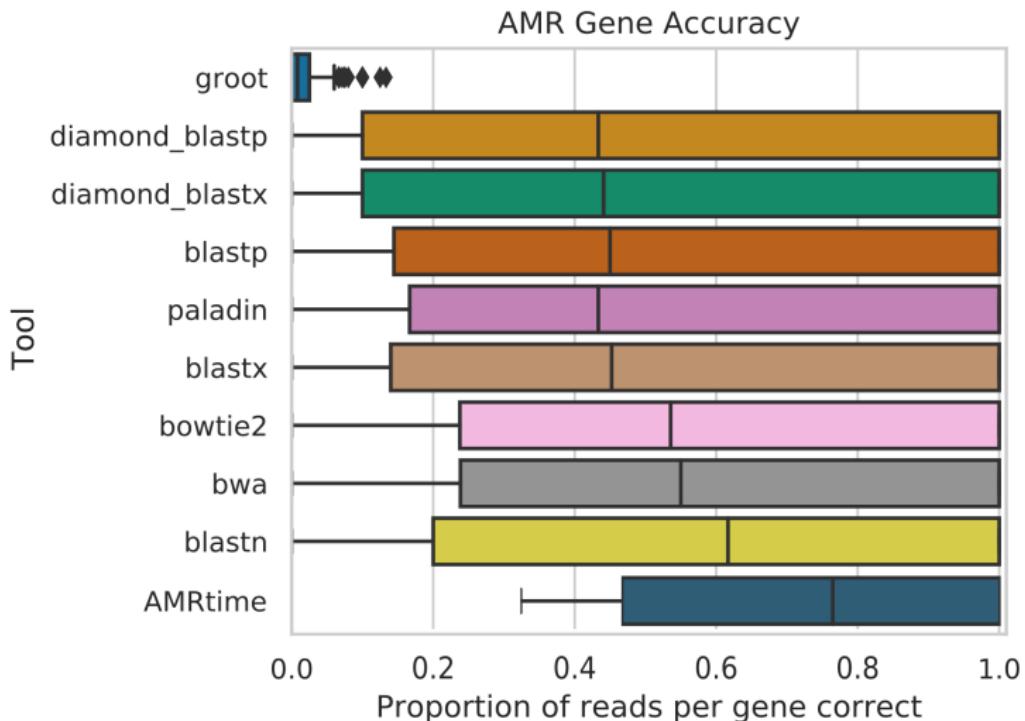


Slightly improved family performance

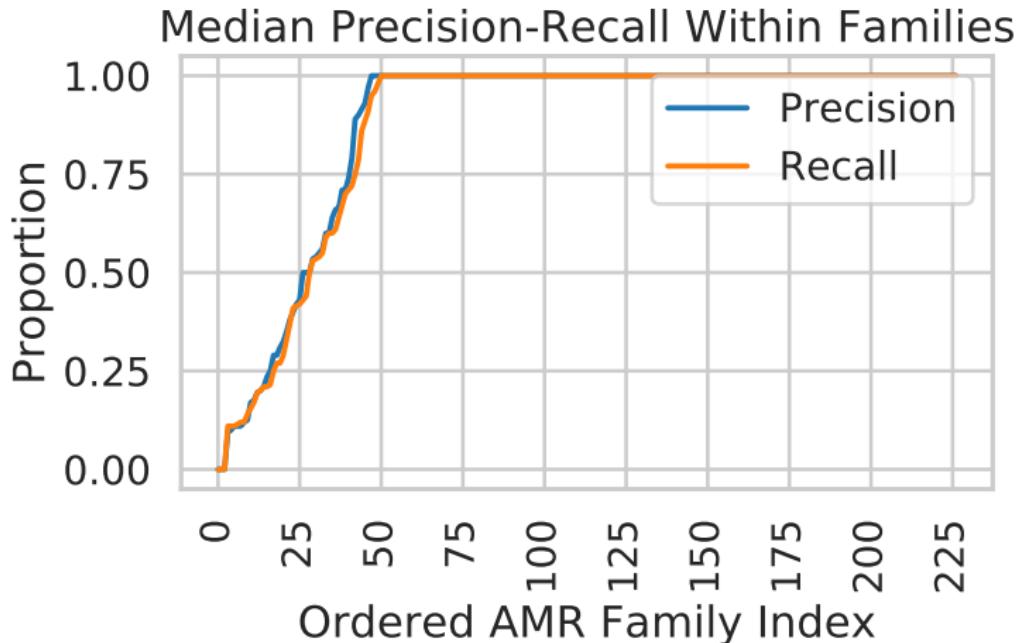


Mean Precision: 0.995, Mean Recall: 0.985

Greatly improved gene performance

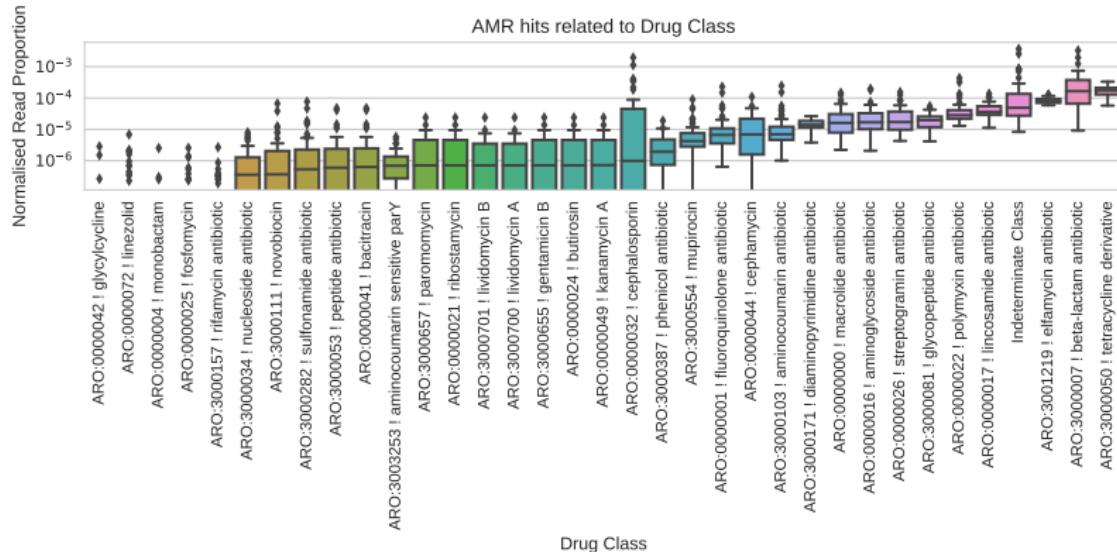


Gains not evenly distributed



- Not enough signal in read so output compatible set
- Some fixed bugs

Metagenomic resistome profile



47 human gut metagenome profiles

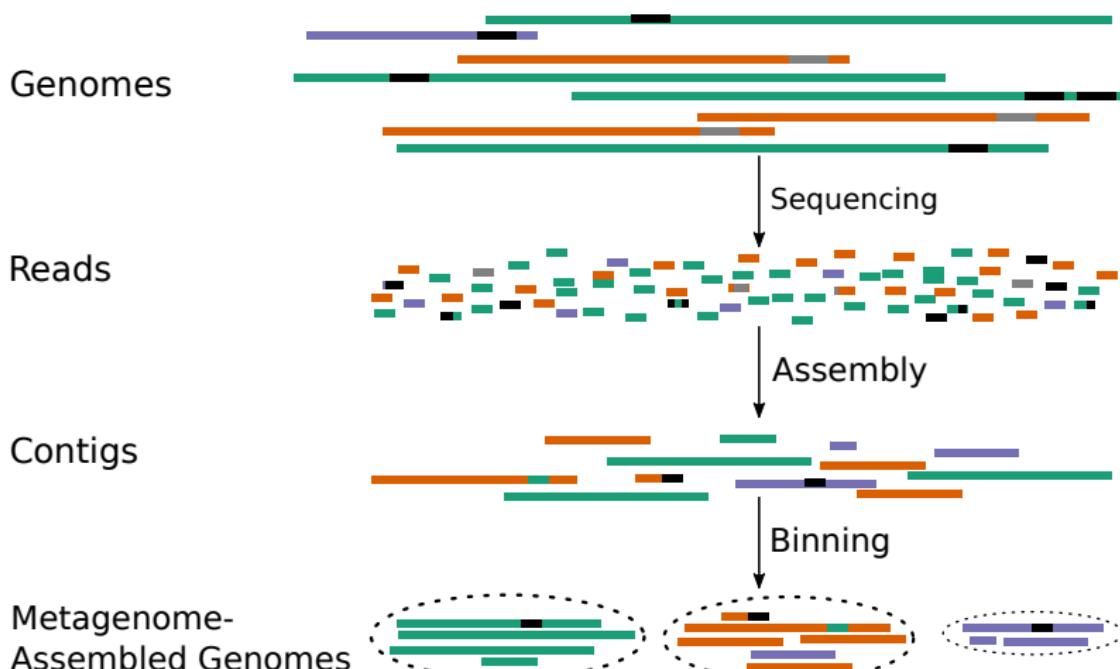
Great, but...

- Known AMR genes
- Is one organism resistant to everything?
- Are many organisms each resistant to one thing?
- Have AMR genes been laterally transferred?

**Can we get the best of
metagenomics and genomics?**

Metagenomic-Assembled Genomes

MAG binning



MAGs are popular

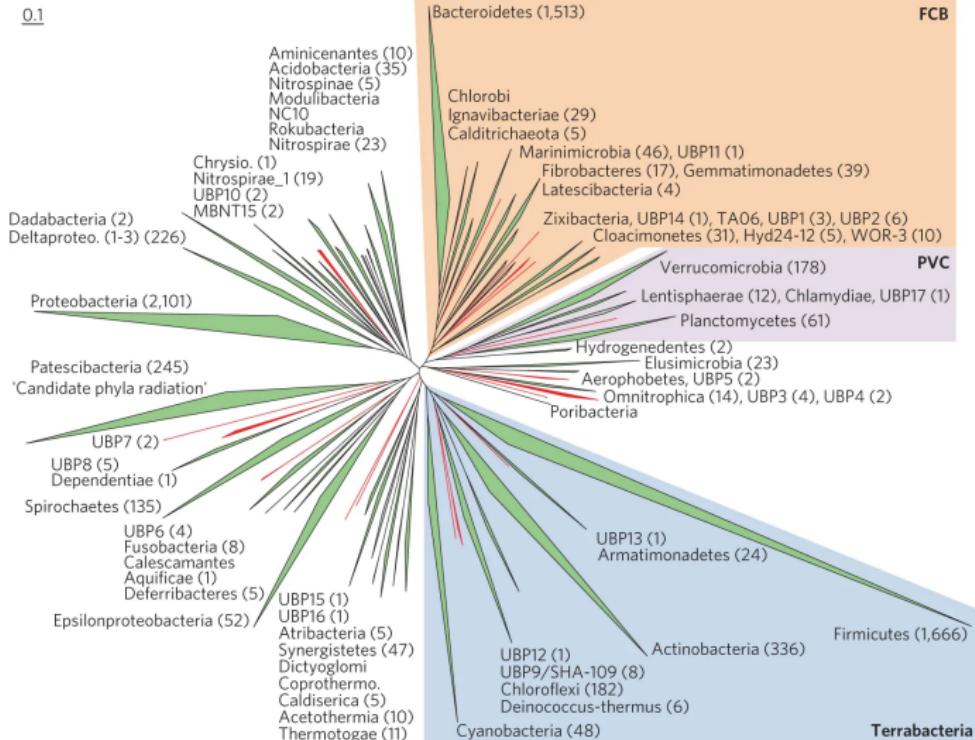


Figure from (Parks et al., 2017)

What about plasmids?

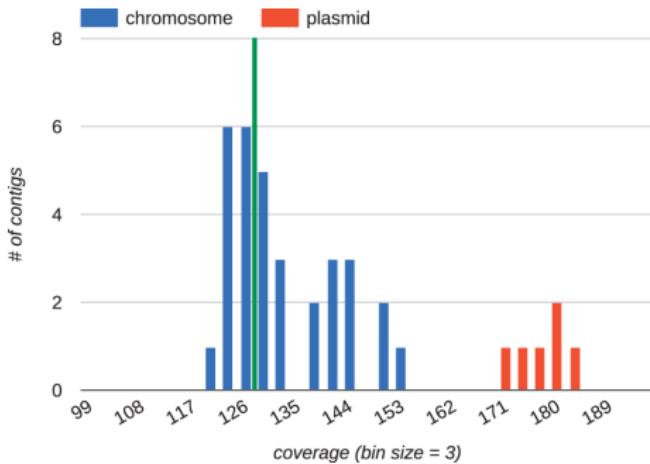
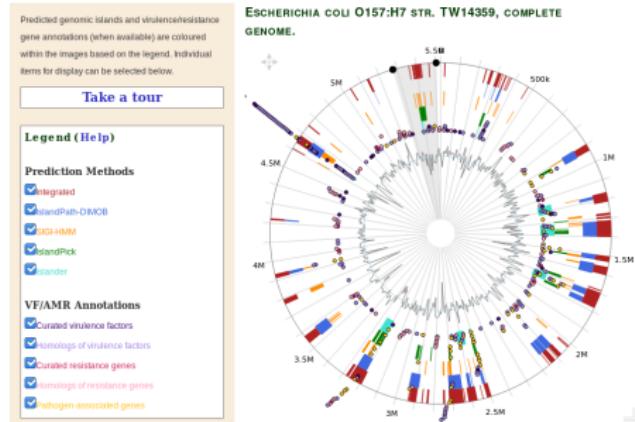


Figure from (Antipov et al., 2016)

- Circular or linear extrachromosomal self-replicating DNA.
- Dissemination of AMR genes.
- Repetitive, variable copy number, different sequence composition.

Or genomic islands



www.pathogenomics.sfu.ca/islandviewer

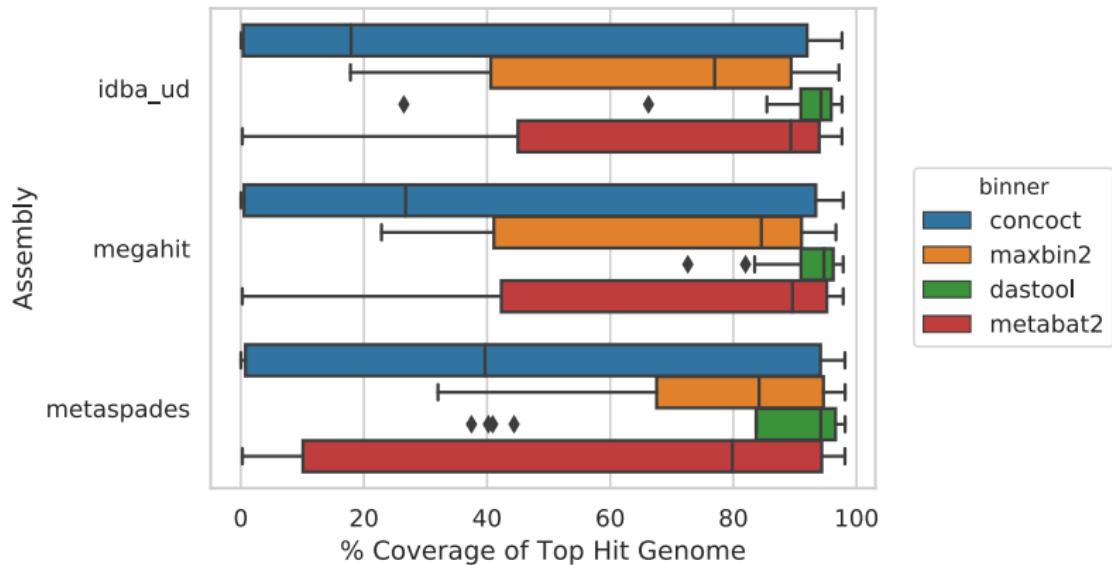
- Clusters of genes acquired through LGT
- Integrons, transposons, integrative and conjugative elements (ICEs) and prophages
- Variable copy number and composition (used by SIGI-HMM, IslandPath-DIMOB)

How well do MAGs recover these sequences?

Time to start simulating again

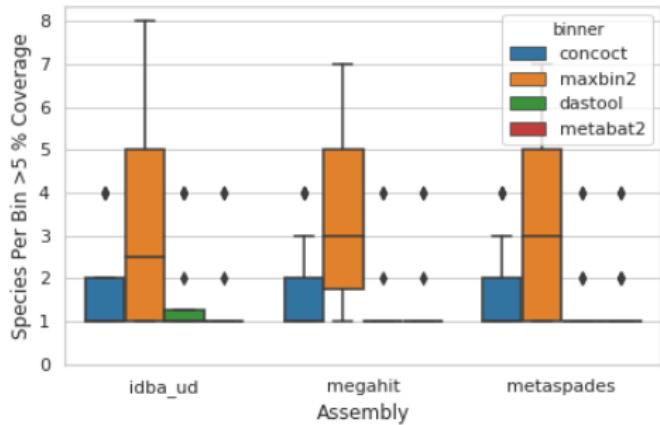
- Simulate some metagenomes (lognormal abundance distribution) from difficult genomes
 - 10 genomes: lots of plasmids
 - 10 genomes: high % of genomic islands (compositional)
 - 10 genomes: low % of genomic islands
- Assembly using 3 alternative methods: IDBA_UD, MetaSPAdes, Megahit
- Bin contigs using 4 different tools: metabat2, maxbin2, concoct, dastool

Chromosomes fairly well binned



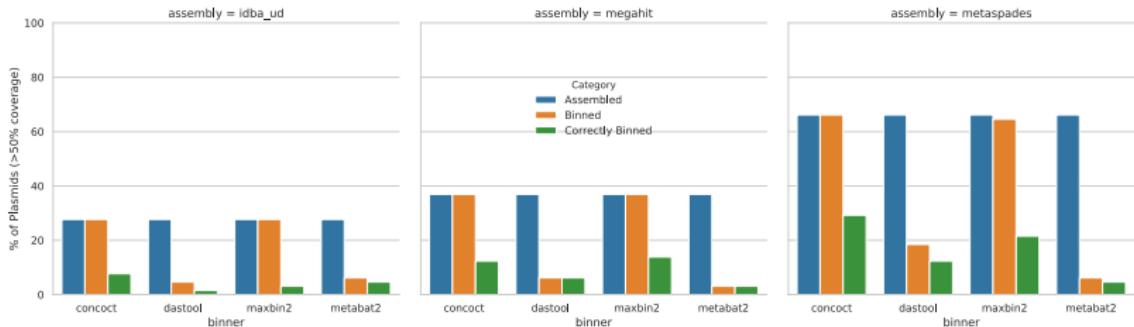
26-94.3% median chromosomal coverage (Pre-print draft
github.com/fmaguire/mag_sim_paper)

Chromosomes fairly well binned



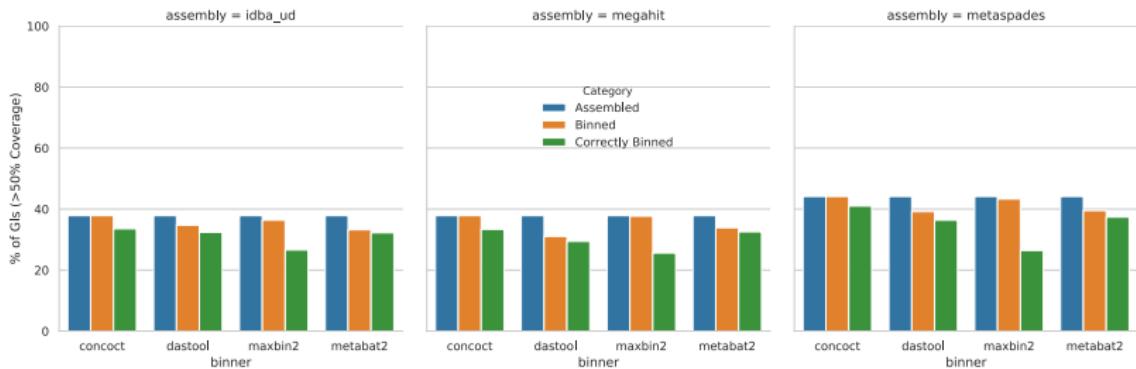
26-94.3% median chromosomal coverage (Pre-print draft
github.com/fmaguire/mag_sim_paper)

Plasmids are not



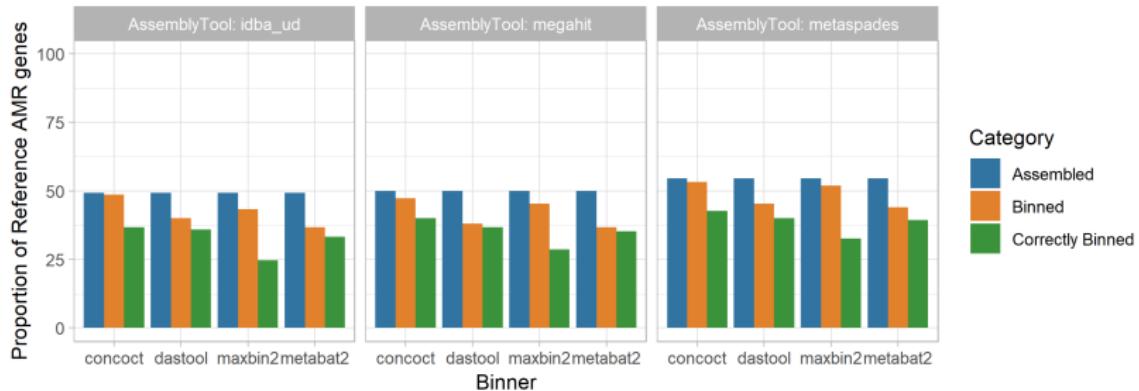
1.5-29.2% plasmids binned

Genomic islands are better but bad



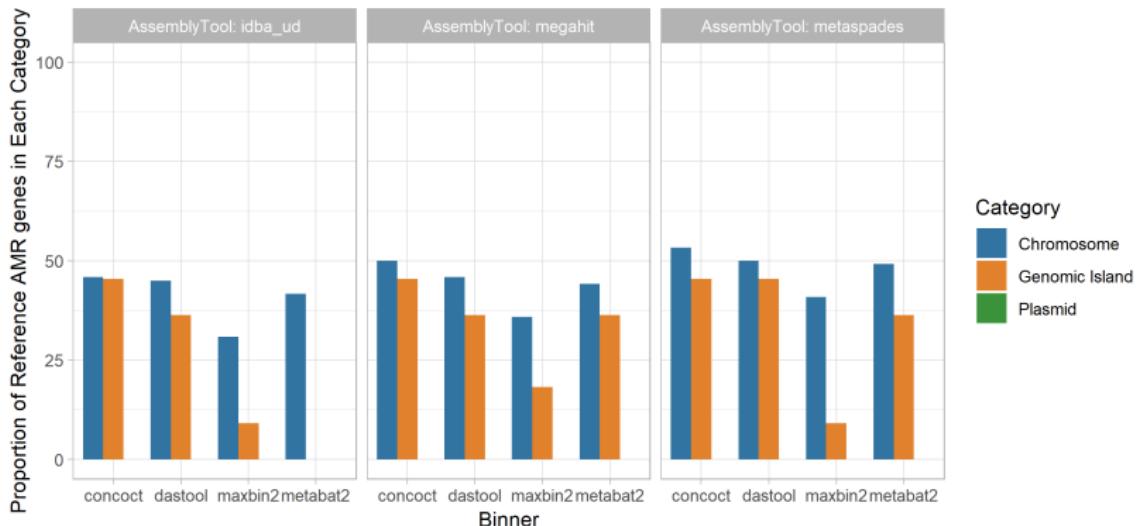
28-42% GIs binned

What about AMR genes?



24-43% AMR genes binned

Which AMR genes are lost?



- 30-53% chromosomal AMR genes (n=120)
- 0-45% genomic island AMR genes (n=11)
- 0% of plasmid AMR genes (n=20)

Be cautious with MAGs

- Regain some context but with biased data loss
- Disproportionate loss of AMR genes
- Mobile Genetic Elements poorly recovered
- Cautionary tale: more processing = more data loss

Conclusions

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Method	Strengths	Weaknesses
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- Simulation fundamental to evaluating approaches
- Characterisation necessary to mitigate weaknesses and promote strengths
- Machine-Learning represents useful tools for this (e.g. AMRtime, gene-free AST prediction models)

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Questions?

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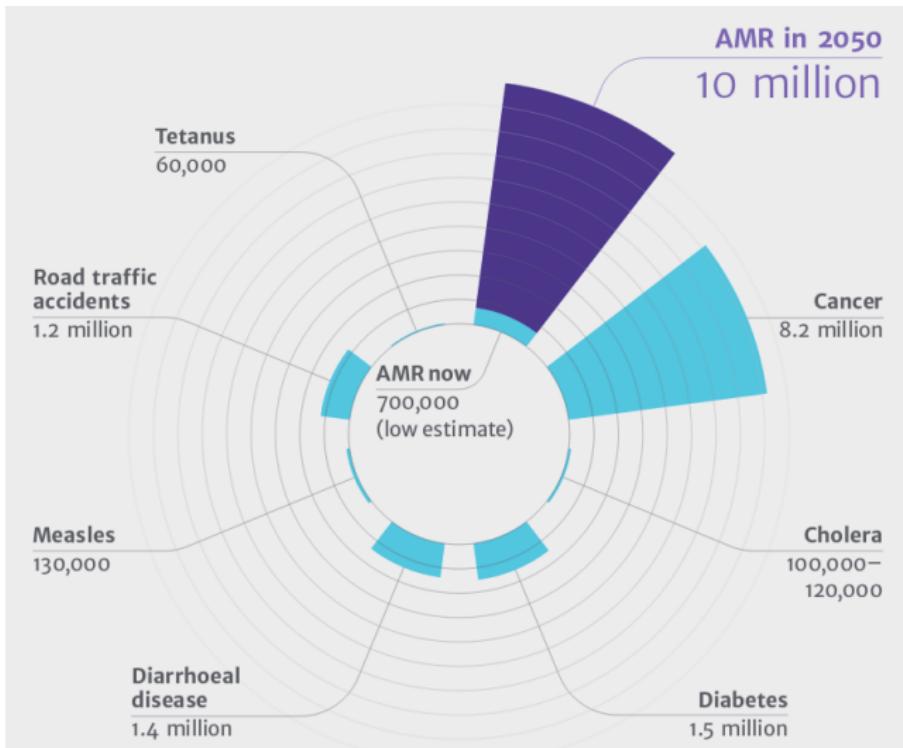
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Backup

10 million deaths?



(on Antimicrobial Resistance, 2016), (de Kraker et al., 2016)

10 million deaths?

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PMCID: PMC5127510
PMID: [27898664](#)

Will 10 Million People Die a Year due to Antimicrobial Resistance by 2050?

[Marlieke E. A. de Kraker](#)^{1,*}, [Andrew J. Stewardson](#)² and [Stephan Harbarth](#)¹

(on Antimicrobial Resistance, 2016), (de Kraker et al., 2016)

Where does 10 million come from?

For 3rd-generation cephalosporin resistant *E. coli*, *K. pneumoniae*, and MRSA:

- Estimate global BSIs (multiply average incidence in tertiary European hospitals by global population).
- Estimate AMR (proportion of resistant blood-cultures per country)
- Extrapolate to other infections sites (via relative incidence to BSI in 2 studies n=16 BSIs)
- Estimate attributable mortality rates from adjusted odds-ratios in an unspecified manner.
- Assume no change in mortality, 40% increase in resistance, and doubled infection rates by 2050.