

Variation Service in PATRIC

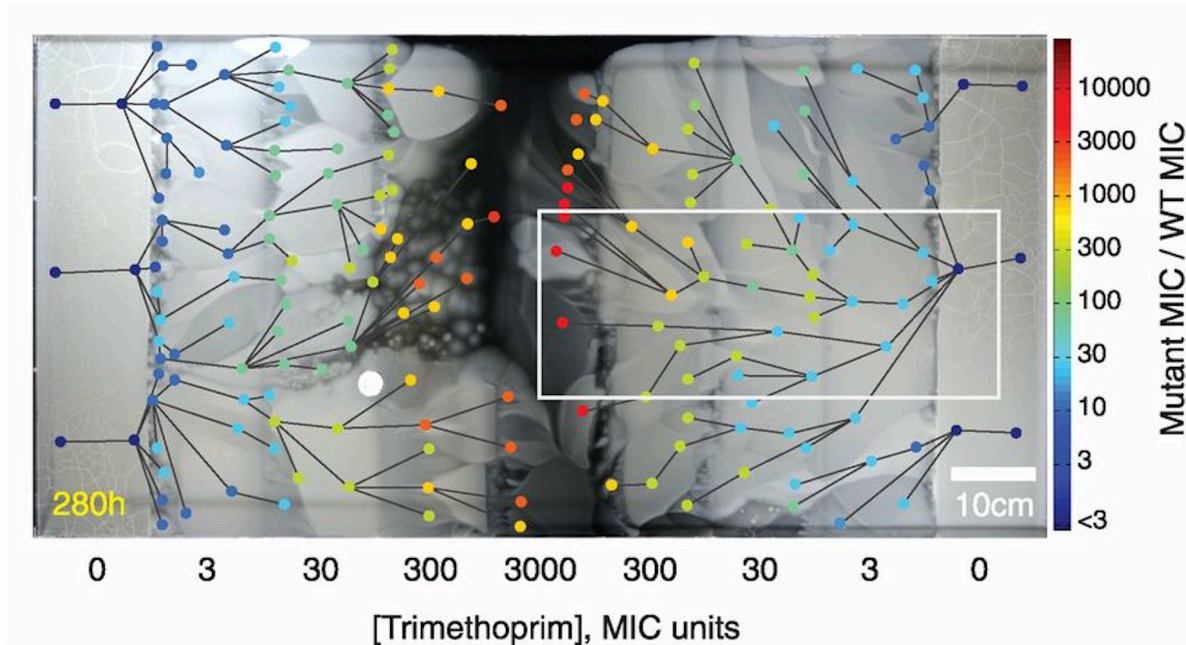
Presented by Fangfang Xia



AMR Evolution in Action



MEGA-plate: Visualizing Antibiotic Landscape



- ▶ Mutated *dnaQ* strains showed increased rates of mutations
- ▶ Mutator phenotype emerged at least six times independently
- ▶ Highly resistant mutants may be trapped
- ▶ Growth restored by compensatory mutations

Spatiotemporal microbial evolution on antibiotic landscapes , Kishony, 2016

Comparative Genomics Goals

- ▶ Identify genetic variations among organisms
- ▶ Associate genetic variations with phenotypes
- ▶ Hypothesize mechanisms that connect genotypes and phenotypes

Genotype-Phenotype Map

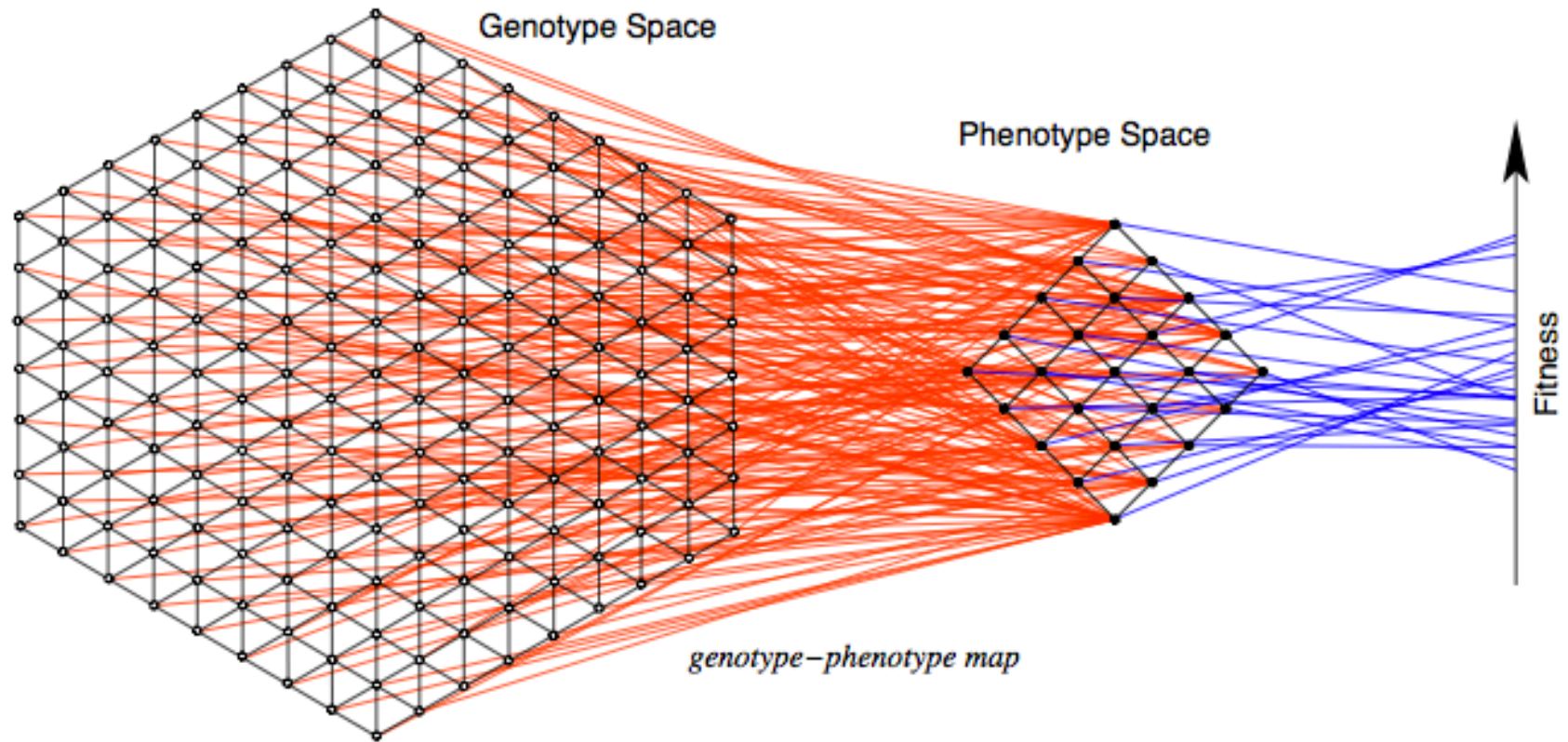
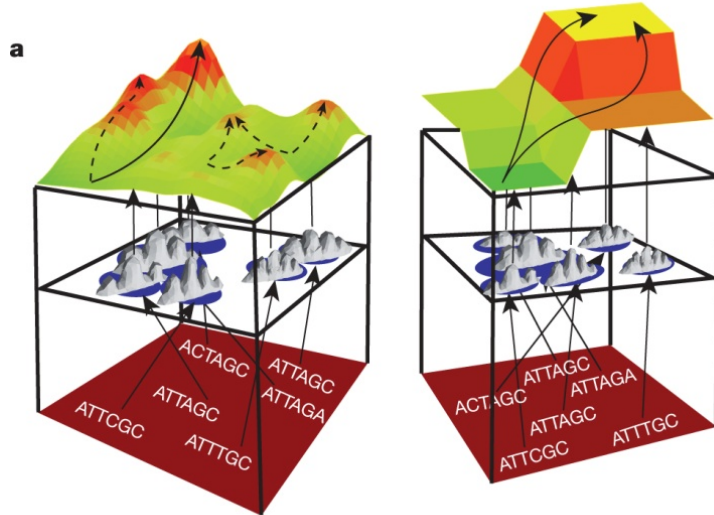


Figure 1. Fitness and Genotype-Phenotype map

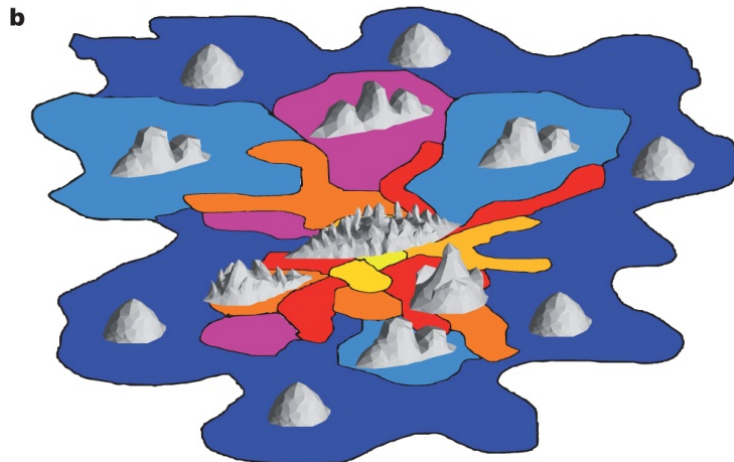
Landscapes and Effective Fitness, Peter Stadler, 2002

Conceptual interpretation of genotype–fitness map



Layers:

- fitness space
- morphological variational space
- genetic variational space



Idealized two–dimensional representation of the tooth model parameter space

Salazar–Ciudad & M Marín–Riera *Nature* 000, 1–4 (2013) doi:10.1038/nature12142

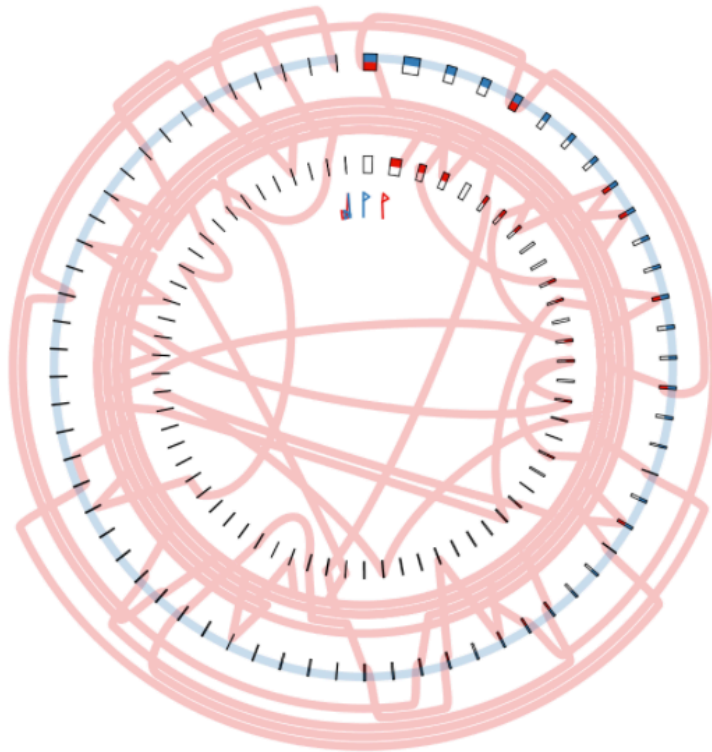
Levels of Genetic Variations

- ▶ Protein family (gains, losses, transfers, CNV)
- ▶ Protein sequence similarity

- ▶ DNA sequence similarity
- ▶ Genome rearrangements
- ▶ DNA local variations
 - SNPs, MNPs, indels, intergenic variations

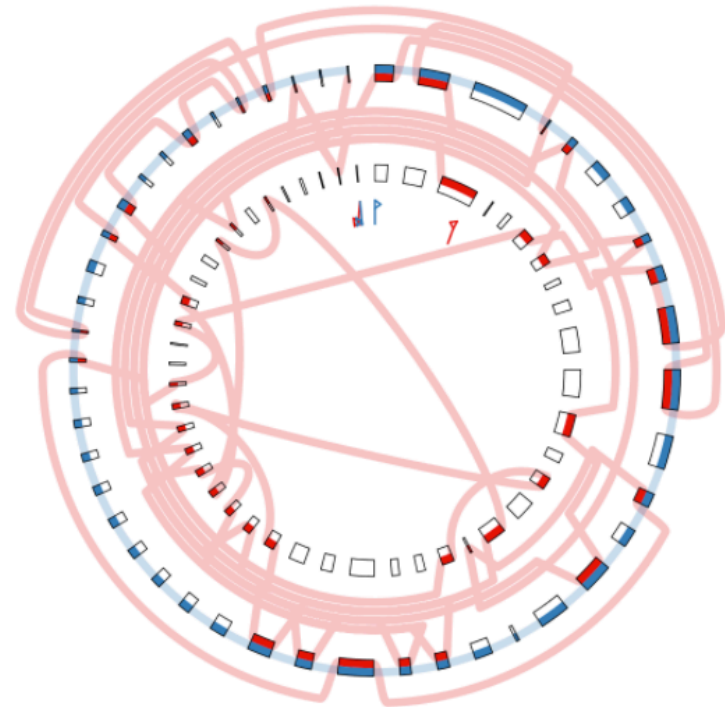
Genome Rearrangement

NNB4-ART-DAR vs NNB4-ART-MS



126 vs 71 contigs

DAR-901 vs MS0193



50 vs 52 contigs

DNA Sequence Similarity

- ▶ Separating genuine sequence difference from artifacts
 - There can be more variations between assemblies of the same strain than different strains

DNA-diff comparison of pairs of assemblies

Sample	Ref \ Qry	D-a6	D-ra	D-sp	M-a6	M-ra	M-sp	SA-M	SB-M	NBB4	H37R
DAR-901_a6_scaffolds	D-a6	0.9998	0.9948	0.9988	0.9955	0.9843	0.9988	0.9987	0.9984	0.3029	0.0419
DAR-901_ray_filtered_scaffolds	D-ra	0.9977	0.9996	0.9992	0.9956	0.9854	0.9993	0.9992	0.9990	0.3058	0.0422
DAR-901_spades_filtered_scaffolds	D-sp	0.9953	0.9927	0.9998	0.9933	0.9825	0.9972	0.9969	0.9969	0.3072	0.0427
MS0193_a6_scaffolds	M-a6	0.9957	0.9928	0.9968	1.0000	0.9825	0.9971	0.9969	0.9967	0.3003	0.0415
MS0193_ray_filtered_scaffolds	M-ra	0.9976	0.9953	0.9992	0.9956	0.9996	0.9994	0.9992	0.9990	0.3077	0.0423
MS0193_spades_filtered_scaffolds	M-sp	0.9975	0.9951	0.9994	0.9958	0.9850	0.9998	0.9993	0.9992	0.3083	0.0427
Subsample_A_MS0193_spades_filtered_scaffolds	SA-M	0.9977	0.9953	0.9995	0.9959	0.9852	0.9996	0.9998	0.9994	0.3083	0.0428
Subsample_B_MS0193_spades_filtered_scaffolds	SB-M	0.9977	0.9953	0.9994	0.9958	0.9850	0.9996	0.9994	0.9998	0.3082	0.0428
Mycobacterium_chubuense_NBB4	NBB4	0.2797	0.2817	0.2832	0.2768	0.2790	0.2833	0.2831	0.2831	1.0000	0.0499
Mycobacterium_tuberculosis_H37Rv	H37R	0.0550	0.0557	0.0560	0.0545	0.0545	0.0560	0.0560	0.0560	0.0692	1.0000

PATRIC Services

PATRIC WORKSHOP @ MICROBE 2017

PATRIC WORKSHOP

PATRIC will be hosting a 1-day workshop entitled "Assemble, Annotate & Analyze Your Own Genome using PATRIC, the All Bacterial Bioinformatics Resource Center" on **number 001-WS** at the ASM Microbes 2017 on June 1, 2017 in New Orleans, LA. The workshop will cover genome assembly and annotation, comparative genomics, RNA-Seq and transcriptomic analysis, and calling SNPs, MNPs, and indels using the Variation pipeline. The workshop will focus on analyzing researcher's private data compared to the available public data. Seating is limited, so please register soon if you are interested. The main ASM Microbes 2017 registration page is [here](#). Note that you must register for the conference in order to sign up for workshops.

REGISTER

- Genomics**
 - Assembly
 - Annotation
 - BLAST
 - Similar Genome Finder
 - Variation Analysis
 - Tn-Seq Analysis
- Protein Tools**
 - Protein Family Sorter
 - Proteome Comparison
- Metabolomics**
 - Comparative Pathway
 - Model Reconstruction
- Data**
 - ID Mapper
- Transcriptomics**
 - Expression Import
 - RNA-Seq Analysis



ASM MICROBE 2017 NEW ORLEANS | ICSB 2017 BLACKSBURG, VA | INVESTIGATING ANTIBIOTIC RESISTANCE | SUPPORTING THE TB COMMUNITY

BROWSE DATA

TOP 10 GENERA



Name	Genomes
Mycobacterium	11644
Streptococcus	11380
Staphylococcus	10633

Variation Job Submission

Variation Analysis

Identify and annotate sequence variation (SNP, MNP, indel) relative to a closely related reference genome.

Paired read library ⓘ ➔

READ FILE 1
SRR3146372_1.fastq.gz ▼ 📁

READ FILE 2
SRR3146372_2.fastq.gz ▼ 📁

Single read library ➔

READ FILE
 ▼ 📁

Parameters ⓘ

Target Genome
Acinetobacter baumannii UH81_364 ▼

OUTPUT FOLDER
Variation Demo ▼ 📁

OUTPUT NAME
Patient_81

Selected libraries ⓘ

Place read files here using the arrow buttons.

P(SRR31..tq.gz, SRR31..tq.gz)	✕
P(SRR31..tq.gz, SRR31..tq.gz)	✕
P(SRR31..tq.gz, SRR31..tq.gz)	✕

Reset Submit

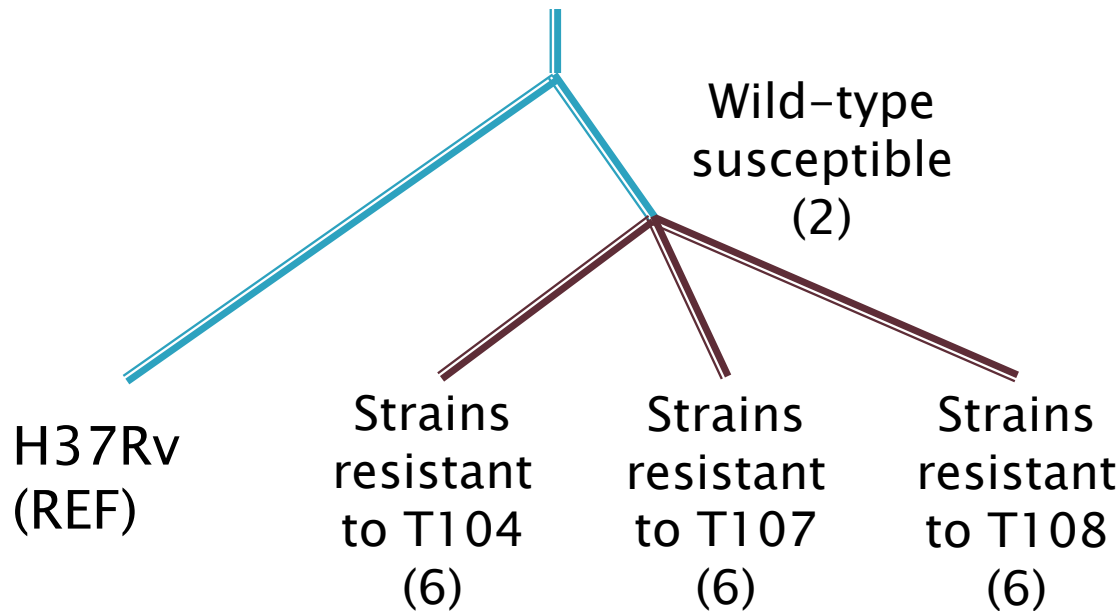
📁 Uploads 0·0 📁 Jobs 46·1·0·6

Variation Analysis

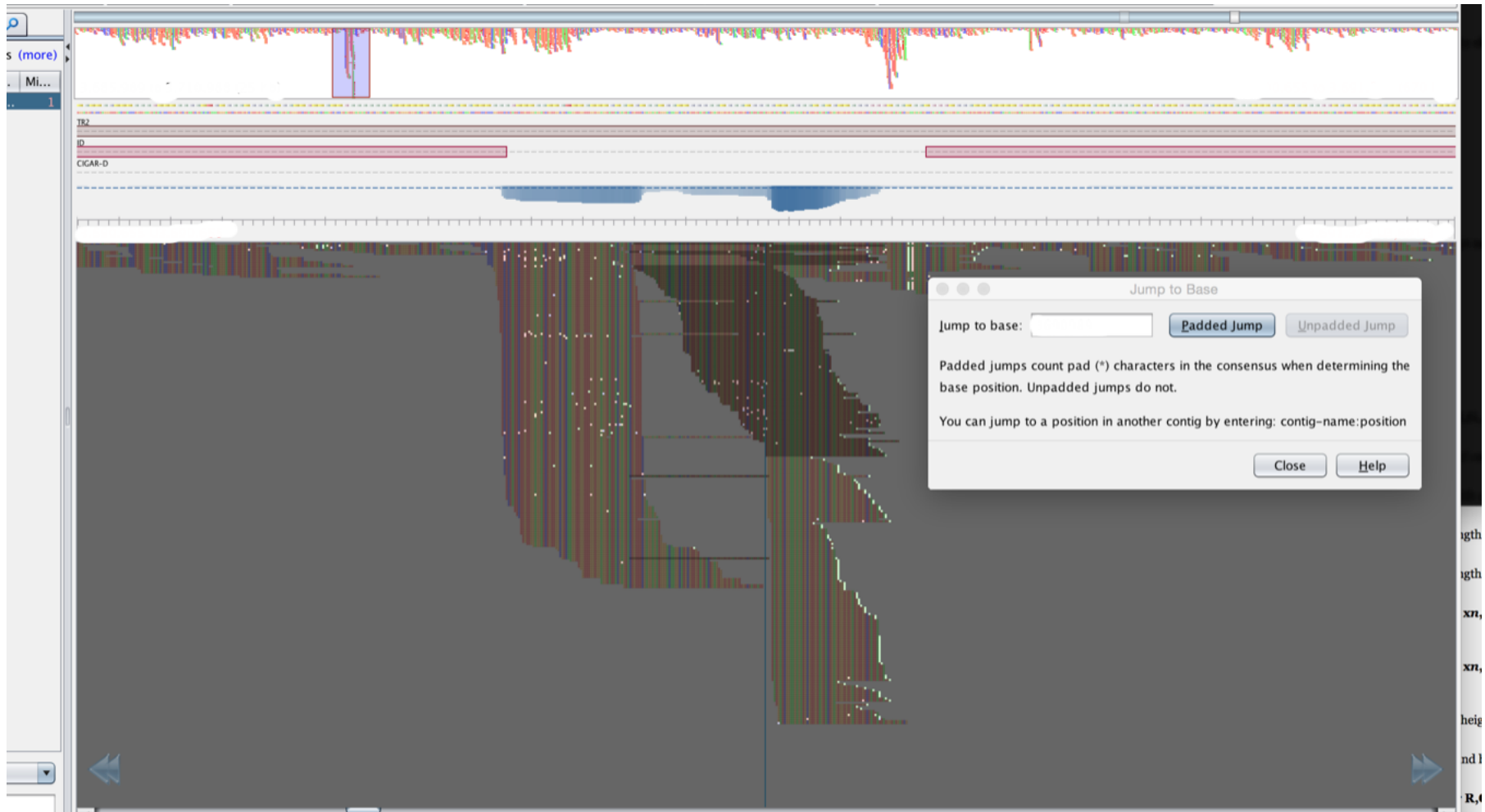
- ▶ Read mapping
- ▶ Coverage statistics
- ▶ Variant calling
- ▶ Variant annotation
- ▶ Cross-sample summary

AMR Analysis with Gyanu Lamichhane Lab

- ▶ 20 genomes
 - 2 are wild-type parent strains
 - 18 offspring strains resistant to 3 carbapenems

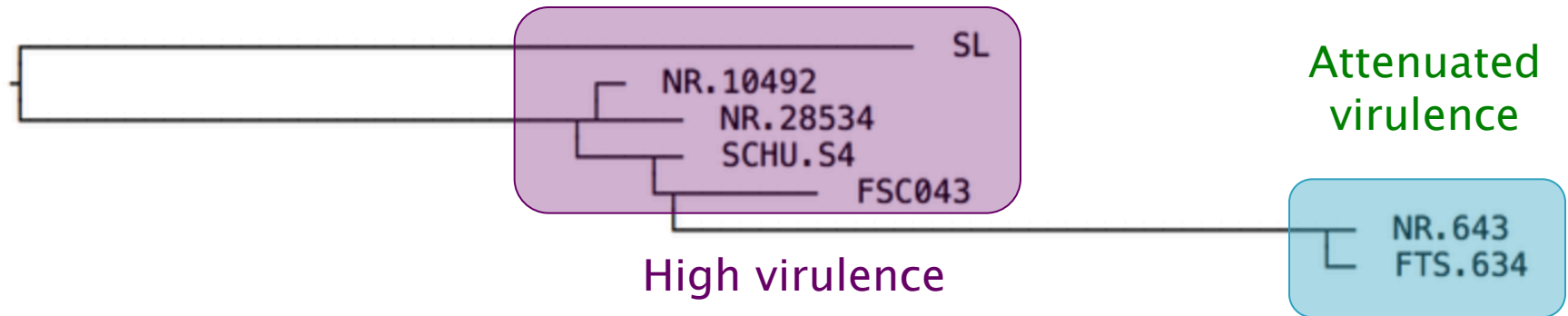


Highly Expressed Intergenic RD



DNA sequence comparison of *Francisella tularensis* Schu S4 substrains

- ▶ Seven substrains in two virulence groups



- ▶ Goal: Identify Regions of Differences
 - Verify the presence of virulence factors
 - Confirm the identity of the organism in stock culture

Separating Genuine Regions of Differences from Assembly Artifacts

AJ749949 REF FSC043 NR-10492 NR-28534 SL FTS-634 NR-643 Type

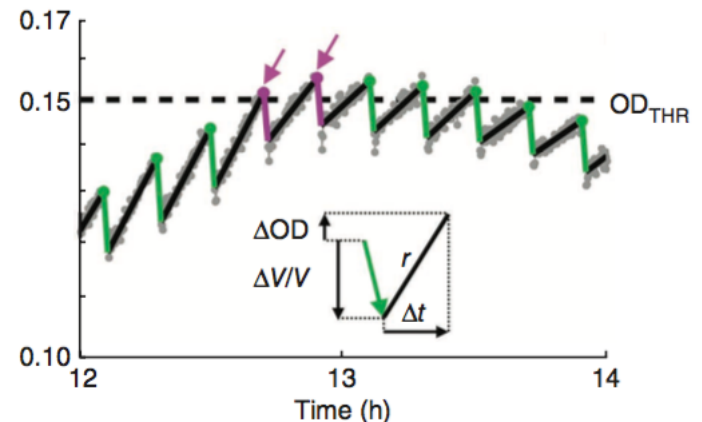
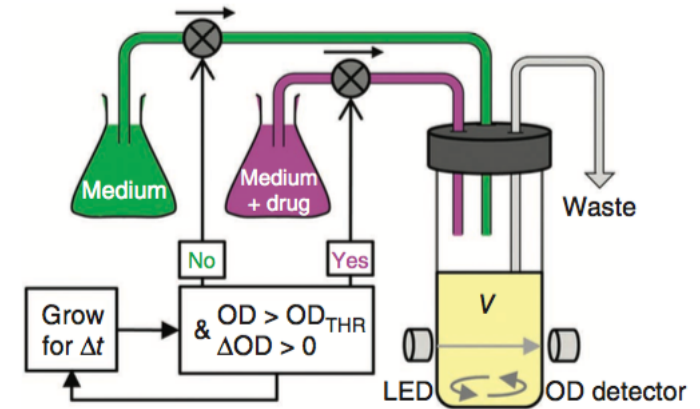
3195*	-	1_0*	-	1_1	C	1_1	C	1_0*	-	1_0*	-	1_0*	-	IND	Repeat region (957 nt, NR-10492_1:1:957) Likely assembly artifacts caused by repeats: The mapped reads that extend from SCHUS4 ref: 13 lines The locus from 3105...4084 has mobile element Of these 50 there are 4 truncated ISFTU1s, so In other words, the reads used for the assembl
3195*	-	1_0*	-	1_2	C	1_2	C	1_0*	-	1_0*	-	1_0*	-	IND	
3195*	-	1_0*	-	1_3	T	1_3	T	1_0*	-	1_0*	-	1_0*	-	IND	
...	*	...	*	*	...	*	...	*	IND	
3195*	-	1_0*	-	1_17	G	1_17	G	1_0*	-	1_0*	-	1_0*	-	IND	
3195*	-	1_0*	-	1_18	A	1_18	A	1_0*	-	1_0*	-	1_0*	-	IND	
3195*	-	1_0*	-	1_19	G	1_19	G	1_0*	-	1_0*	-	1_0*	-	IND	
3195*	-	1_0*	-	1_20	T	1_20	T	1_1	T	1_0*	-	1_0*	-	IND	
3195*	-	1_0*	-	1_21	A	1_21	A	1_2	A	1_0*	-	1_0*	-	IND	
3195*	-	1_0*	-	1_22	T	1_22	T	1_3	T	1_0*	-	1_0*	-	IND	
...	*	...	*	*	...	*	...	*	IND	
3195*	-	1_0*	-	1_93	T	1_93	T	1_74	T	1_0*	-	1_0*	-	IND	
3195*	-	1_0*	-	1_94	T	1_94	T	1_75	T	1_0*	-	1_0*	-	IND	
3195*	-	1_0*	-	1_95	A	1_95	A	1_76	A	1_0*	-	1_0*	-	IND	
3196	A	1_0*	-	1_96	A	1_96	A	1_77	A	1_0*	-	1_0*	-	IND	
3197	A	1_0*	-	1_97	A	1_97	A	1_78	A	1_0*	-	1_0*	-	IND	
3198	G	1_0*	-	1_98	G	1_98	G	1_79	G	1_0*	-	1_0*	-	IND	

1419874	A	1_1417461	A	1_1418732	A	1_1416876	A	1_1417139	A	1_1416598	A	1_1416737	A		
1419875	T	1_1417462	T	1_1418733	T	1_1416877	T	1_1417140	T	1_1416599	T	1_1416738	T		
1419876	G	1_1417463	G	1_1418734	G	1_1416878	G	1_1417141	G	1_1416600	G	1_1416739	G		
1419877	C	1_1417464	C	1_1418735	C	1_1416879	C	1_1417142	C	1_1416601	T	1_1416740	T	SNP	Nonsyn (Cct -> Tct, P -> S);
1419878	C	1_1417465	C	1_1418736	C	1_1416880	C	1_1417143	C	1_1416602	C	1_1416741	C		
1419879	T	1_1417466	T	1_1418737	T	1_1416881	T	1_1417144	T	1_1416603	T	1_1416742	T		
1419880	A	1_1417467	A	1_1418738	A	1_1416882	A	1_1417145	A	1_1416604	A	1_1416743	A		
1423159	C	1_1420746	C	1_1422017	C	1_1420161	C	1_1420424	C	1_1419883	C	1_1420022	C		
1423160	A	1_1420747	A	1_1422018	A	1_1420162	A	1_1420425	A	1_1419884	A	1_1420023	A		
1423161	C	1_1420748	C	1_1422019	C	1_1420163	C	1_1420426	C	1_1419885	C	1_1420024	C		
1423162	A	1_1420749	A	1_1422020	A	1_1420164	A	1_1420427	A	1_1419886	G	1_1420025	G	SNP	Nonsyn (Agt -> Ggt, S -> G);
1423163	G	1_1420750	G	1_1422021	G	1_1420165	G	1_1420428	G	1_1419887	G	1_1420026	G		
1423164	T	1_1420751	T	1_1422022	T	1_1420166	T	1_1420429	T	1_1419888	T	1_1420027	T		
1423165	A	1_1420752	A	1_1422023	A	1_1420167	A	1_1420430	A	1_1419889	A	1_1420028	A		

Evolutionary paths to antibiotic resistance under dynamically sustained drug selection





Erdal Toprak^{1,6}, Adrian Veres^{2,6}, Jean-Baptiste Michel^{1,3}, Remy Chait¹, Daniel L Hartl⁴ & Roy Kishony^{1,5}

Antibiotic resistance can evolve through the sequential accumulation of multiple mutations¹. To study such gradual evolution, we developed a selection device, the 'morbidostat', that continuously monitors bacterial growth and dynamically regulates drug concentrations, such that the evolving population is constantly challenged²⁻⁵. We analyzed the evolution of resistance in *Escherichia coli* under selection with single drugs, including chloramphenicol, doxycycline and trimethoprim. Over a period of ~20 days, resistance levels increased dramatically, with parallel populations showing similar phenotypic trajectories. Whole-genome sequencing of the evolved strains identified mutations both specific to resistance to a particular drug and shared in resistance to multiple drugs. Chloramphenicol and doxycycline resistance evolved smoothly through diverse combinations of mutations in genes involved in translation, transcription and transport³. In contrast, trimethoprim resistance evolved in a stepwise manner^{1,6}, through mutations restricted to the gene encoding the enzyme dihydrofolate reductase (*DHFR*)^{7,8}. Sequencing of *DHFR* over the time course of the experiment showed that parallel populations evolved similar mutations and acquired them in a similar order⁹.



Demo Data

- ▶ Reference genome:
 - Escherichia coli str. K-12 substr. MG1655
- ▶ Reads:
 - BioSample: SAMN00723035

 SRR396518.fastq.gz	237 MB	7/26/16, 3:57:00 AM
 SRR396519.fastq.gz	264 MB	7/26/16, 4:00:00 AM
 SRR396520.fastq.gz	218 MB	7/26/16, 4:02:00 AM
 SRR396521.fastq.gz	271 MB	7/26/16, 4:05:00 AM

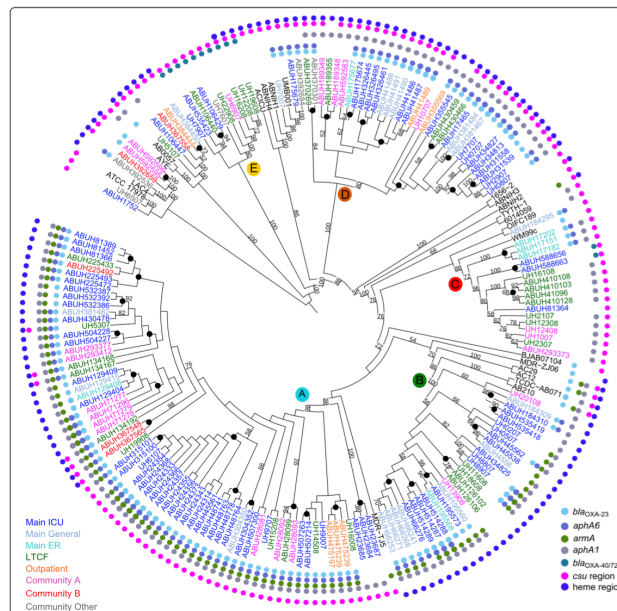
RESEARCH

Open Access

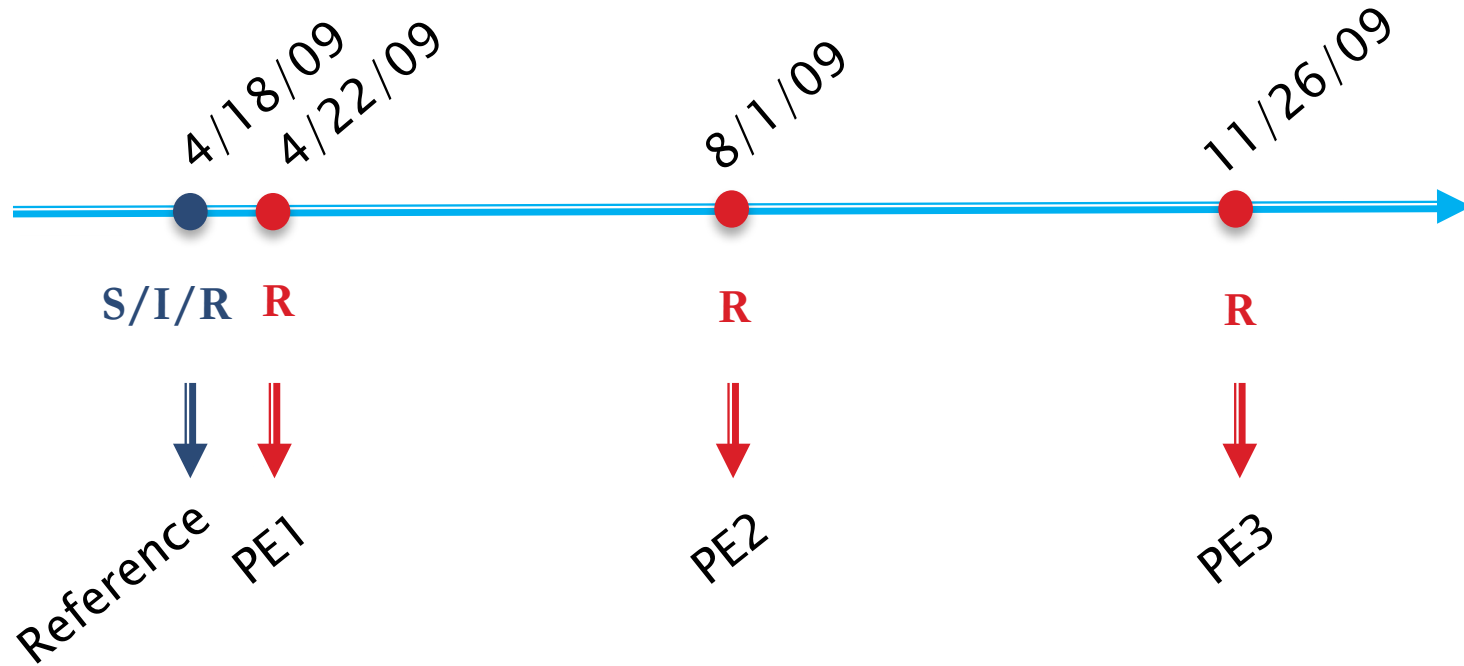


Genome dynamics of multidrug-resistant *Acinetobacter baumannii* during infection and treatment

Meredith S. Wright¹, Alina Iovleva², Michael R. Jacobs^{3,4}, Robert A. Bonomo^{5,6} and Mark D. Adams^{1*}



Serial Samples from Patient 81



Results

▶ Summary

- A total of 552 variants have been found. 293 of these variants are identified in all read libraries.

▶ Questions

- Which RDs are significant?
 - frameshift, stop gain/loss, nonsyn, regulatory region
 - high fraction, high coverage
- Which RDs are common to the later samples?

Inferred Variations by Sample

[1] PE1 (SRR3146360_1.fastq.gz,SRR3146360_2.fastq.gz)

Total reads	=	12261074
Properly mapped reads	=	11919546 (97.21%)
Total reference bases	=	3977313
Median base coverage	=	18
Mean base coverage	=	63.4
Bases with zero coverage	=	462022 (11.616%)
Bases with <=10 coverage	=	1606088 (40.381%), in 111711 contiguous regions
Raw FreeBayes variants	=	2056
High quality variants	=	719

[2] PE2 (SRR3146370_1.fastq.gz,SRR3146370_2.fastq.gz)

Total reads	=	10282737
Properly mapped reads	=	9865157 (95.94%)
Total reference bases	=	3976580
Median base coverage	=	18
Mean base coverage	=	67.1
Bases with zero coverage	=	436258 (10.971%)
Bases with <=10 coverage	=	1577711 (39.675%), in 114191 contiguous regions
Raw FreeBayes variants	=	2027
High quality variants	=	693

Comparison with Paper Results

- ▶ *Phylogenetic analysis revealed that a significant fraction of apparently persistent infections are in fact due to re-infection with new strains. SNVs primarily resulted in protein coding changes, and IS events primarily interrupted genes or were in an orientation such that the adjacent gene would be over-expressed. **Mutations acquired during infection were over-represented in transcriptional regulators, notably pmrAB and adeRS, which can mediate resistance to the last line therapies colistin and tigecycline, respectively, as well as transporters, surface structures, and iron acquisition genes.***

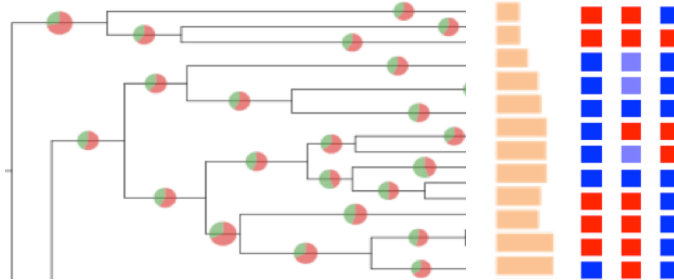
Feature Space to Compute Against

- ▶ Genetic variations
 - Protein presence/absence
 - SNPs, MNPs
 - Insertions/deletions
 - Genome rearrangements
 - Kmer counts
 - Evolutionary histories
- ▶ Machine learning
- ▶ Statistics
- ▶ Omics

Bacterial-Genome Association Studies

Genotypic Variation

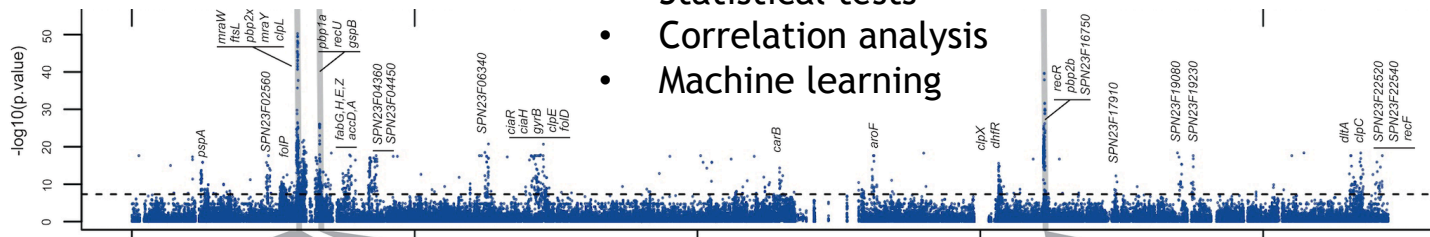
- What's changed:
 - gene gains/losses
- SNPs, insertions, deletions
- Genome rearrangements
- CNV in repeat regions



Phenotypic Variation

- Antibiotic resistance
- Differential growth rates on media types
- Survival/growth under different conditions
- Metabolomic fingerprints

Genotype-phenotype association



- Statistical tests
- Correlation analysis
- Machine learning

Parkhill et al, 2014

Curated subsystems

Specialty gene databases

Transcriptome analysis

Metabolic and regulatory models

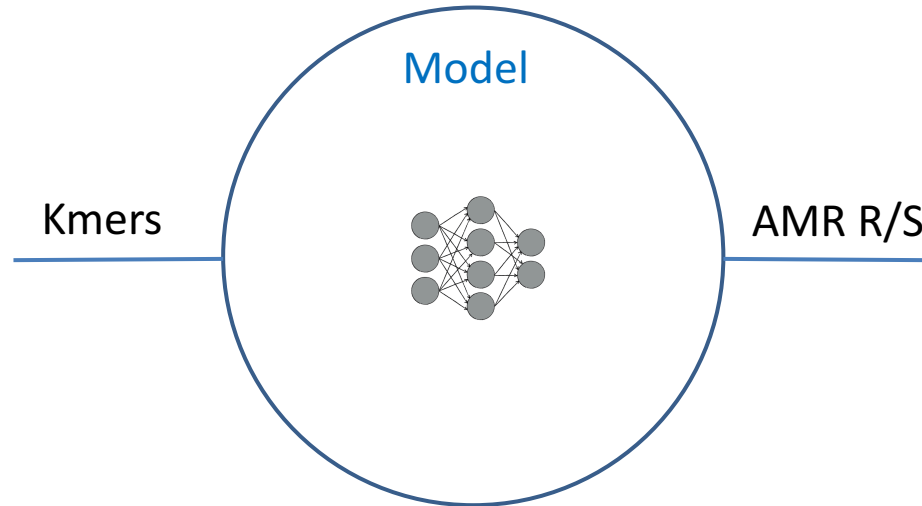
Protein – protein association graphs

Experimental validation

Mechanism elucidation

Deep Learning for Phenotype Prediction

Input



Output

Protein families
Gene clusters
Operons
Expression
Interactions
Motility
Life style
Metabolites
AMR
Virulence
Taxonomy
Functions
Pathways

Sequence

