

Rational Design of Antibiotic Treatment Plans

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Joint paper with Portia Mira, Kristina Crona,
Devin Greene, Juan Meza and Miriam Barlow

PLOS One, May 6, 2015

Our Abstract

“The development of reliable methods for restoring susceptibility after antibiotic resistance arises has proven elusive. A greater understanding of the relationship between antibiotic administration and the evolution of resistance is key to overcoming this challenge.

*Here we present a data-driven mathematical approach for developing **antibiotic treatment plans** that can **reverse the evolution** of antibiotic resistance determinants.*

*We have generated adaptive landscapes for 16 genotypes of the **TEM β -lactamase** that vary from the wild type genotype TEM-1 through all combinations of four amino acid substitutions*

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*Overall this study shows that there is promise for **reversing the evolution of resistance through antibiotic treatment plans.**”*

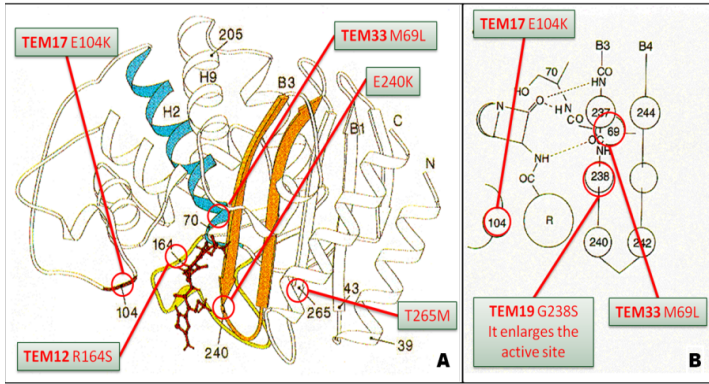
Time Machine



Think of **reversing the evolution** as traveling back in time.

Chemistry

Our time traveler is the **resistance gene** TEM β -lactamase. Identified by Naomi Datta in 1963; named after patient Temoneira.



Wiki: Beta-lactamases are enzymes produced by some bacteria that provide resistance to ... [antibiotics like penicillins](#), cephamycins, ... Beta-lactamase provides antibiotic resistance by breaking the antibiotics' structure. These antibiotics all have a common element in their molecular structure: a four-atom ring known as a β -lactam ...

Evolution Observed

Of 190 clinically identified TEM genotypes, 174 differ from the wild type TEM-1 by at most four amino acid substitutions:

Number of amino acid substitutions	Number of TEM enzymes
1	53
2	53
3	37
4	31
5	10
6	2
7	2
8	0
9	0
10	1
11	1

The Barlow lab studies a system with four amino acid substitutions.

TEM-50 system

The **TEM-50 genotype** is highly evolved.

It differs from the TEM-1 in four amino acid substitutions:



Experiments: The Barlow lab

- ▶ created all 16 genotypes between TEM-1 and TEM-50
- ▶ measured the growth rates of 12 replicates of E.coli DH5 α -E
- ▶ repeated in the presence of each of 15 β -lactam antibiotics

AMP, AM, CEC, CTX, ZOX, CXM, CRO, AMC, CAZ, CTT, SAM, CPR, CPD, TZP, FEP

The 4-dimensional cube

Mutant	Isolated	Binary Allele Code
Wild Type	TEM-1	0000
M69L	TEM-33	1000
E104K	TEM-17	0100
G238S	TEM-19	0010
N276D	TEM-84	0001
M69L/E104K	-	1100
M69L/G238S	-	1010
M69L/N276D	TEM-35	1001
G238S/E104K	TEM-15	0110
G238S/N276D	-	0011
N276D/E104K	-	0101
M69L/E104K/ N276D	-	1101
M69L/E104K/ G238S	-	1110
G238S/N276D/ E104K	-	0111
G238S/N276D/ M69L	-	1011
TEM-50	TEM-50	1111

The data

Measure the fitness of each of the 16 genotypes

Treatment (with 1 antibiotic)												Control (without antibiotic)													
		1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24
0000	A	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24
0100	B	25	26	27	28	29	30	31	32	33	34	35	36	37	38	39	40	41	42	43	44	45	46	47	48
0100	C	49	50	51	52	53	54	55	56	57	58	59	60	61	62	63	64	65	66	67	68	69	70	71	72
0010	D	73	74	75	76	77	78	79	80	81	82	83	84	85	86	87	88	89	90	91	92	93	94	95	96
0001	E	97	98	99	100	101	102	103	104	105	106	107	108	109	110	111	112	113	114	115	116	117	118	119	120
1100	F	121	122	123	124	125	126	127	128	129	130	131	132	133	134	135	136	137	138	139	140	141	142	143	144
1010	G	145	146	147	148	149	150	151	152	153	154	155	156	157	158	159	160	161	162	163	164	165	166	167	168
1001	H	169	170	171	172	173	174	175	176	177	178	179	180	181	182	183	184	185	186	187	188	189	190	191	192
0110	I	193	194	195	196	197	198	199	200	201	202	203	204	205	206	207	208	209	210	211	212	213	214	215	216
0101	J	217	218	219	220	221	222	223	224	225	226	227	228	229	230	231	232	233	234	235	236	237	238	239	240
0011	K	241	242	243	244	245	246	247	248	249	250	251	252	253	254	255	256	257	258	259	260	261	262	263	264
1011	L	265	266	267	268	269	270	271	272	273	274	275	276	277	278	279	280	281	282	283	284	285	286	287	288
1101	M	289	290	291	292	293	294	295	296	297	298	299	300	301	302	303	304	305	306	307	308	309	310	311	312
0111	N	313	314	315	316	317	318	319	320	321	322	323	324	325	326	327	328	329	330	331	332	333	334	335	336
1011	O	337	338	339	340	341	342	343	344	345	346	347	348	349	350	351	352	353	354	355	356	357	358	359	360
1111	P	361	362	363	364	365	366	367	368	369	370	371	372	373	374	375	376	377	378	379	380	381	382	383	384

... in the presence of each of 15 antibiotics.

Fitness landscapes

For each of the 15 **antibiotics** AMP, AM, CEC, CTX, ZOX, CXM, CRO, AMC, CAZ, CTT, SAM, CPR, CPD, TZP, FEP we have a

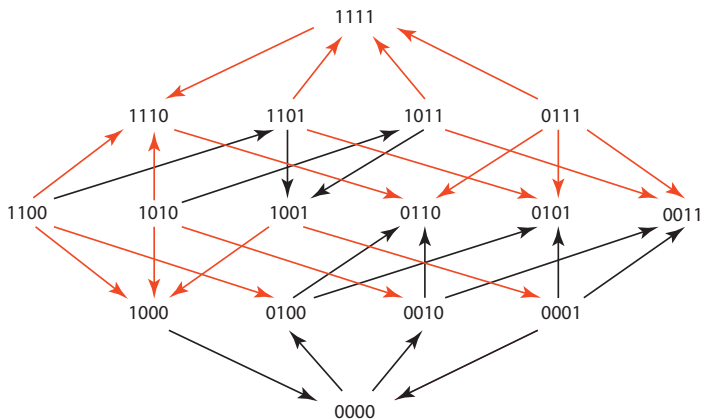
$2 \times 2 \times 2 \times 2$ -table $\mathbf{f} = (f_{ijkl})$ with entries in \mathbb{R}

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We care about the **dynamics** on such fitness landscapes:



Population genetics

A *substitution model* is a function

$$M : \mathbb{R}^{16} \longrightarrow \mathbb{R}^{16 \times 16}$$

that assigns to each fitness landscape \mathbf{f} a transition matrix $M(\mathbf{f})$.

We use two simple models:

- ▶ The **Equal Probability Model** (EPM)

$$M_{uv} = 1/k$$

if $u \rightarrow v$ and genotype u has k outgoing arrows.

- ▶ The **Correlated Probability Model** (CPM)

$$M_{uv} = \frac{f_v - f_u}{\sum_w (f_w - f_u)}$$

if $u \rightarrow v$ and the sum is over all genotypes w with $u \rightarrow w$.

Data yield matrices $M^{XPM}(\mathbf{f}_\star)$ of format 16×16 , where $X = E, C$ and $\star = \text{AMP, AM, CEC, CTX, ZOX, CXM, CRO, AMC, CAZ, CTT, SAM, CPR, CPD, TZP, FEP}$

Matrix algebra

Fix EPM or CPM, and consider the matrix product

$$M(\mathbf{f}_{a_1}) \cdot M(\mathbf{f}_{a_2}) \cdot M(\mathbf{f}_{a_3}) \cdots M(\mathbf{f}_{a_k})$$

where $a_1, a_2, a_3, \dots, a_k$ is any sequence of antibiotics.

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There are 15^k such sequences!

Which ones should the doctors try?

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Which ones should the doctors try?

Fix two genotypes $u, v \in \{0, 1\}^4$.

The entry of this matrix in row u and column v is the fixation probability of genotype u mutating to genotype v .

Combinatorial optimization

Let $v = \text{TEM-1}$ and u any other genotype.

Time Machine Problem: Maximize the (u, v) entry of the matrix

$$M(\mathbf{f}_{a_1}) \cdot M(\mathbf{f}_{a_2}) \cdot M(\mathbf{f}_{a_3}) \cdots M(\mathbf{f}_{a_k})$$

over all 15^k antibiotic sequences of length k .

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Idea: Travel back in time to 1963



Just do it

Time Machine Problem: Given 15 transition matrices of format 16×16 , consider all products of length k , and maximize a particular entry in the first row.

We implemented a brute-force approach for $k = 2, \dots, 6$ in maple, and we computed optimal solutions for both substitution models.

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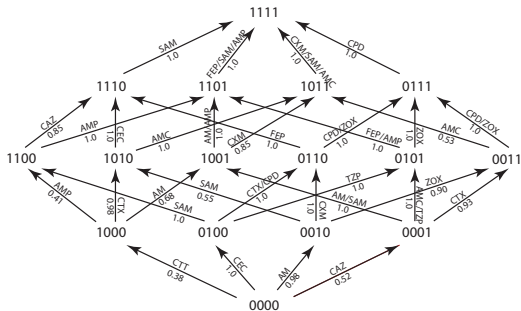
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Results:

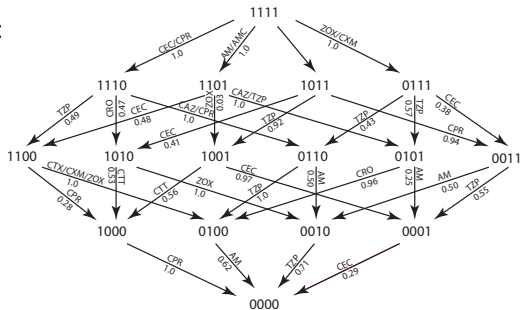
Starting Genotype	1 Step	#	2 Step	#	3 Step	#	4 Step	#	5 Step	#
1000	1.0	1	1.0	3	1.0	7	1.0	15	1.0	31
0100	0.617	1	0.617	6	0.617	36	0.617	219	0.617	1360
0010	0.714	1	0.714	2	0.714	3	0.714	4	0.714	5
0001	0.287	1	0.287	1	0.592	2	0.592	18	0.726	2
1100			0.617	3	0.617	18	0.617	108	0.617	657
1010			0.715	1	0.715	6	0.715	27	0.715	112
1001			0.559	1	0.559	4	0.726	1	0.726	2
0110			0.617	1	0.617	10	0.617	78	0.617	555
0101			0.592	1	0.592	9	0.612	1	0.612	9
0011			0.361	1	0.361	9	0.586	1	0.599	2
1110			-		0.617	2	0.617	24	0.617	215
1101			-		0.592	2	0.592	24	0.617	12
1011			-		0.532	1	0.532	1	0.684	1
0111			-		0.586	1	0.600	1	0.617	4
1111			-		-		0.617	4	0.617	72

Biomedical interpretation

Since 1963:



Back to wild type:



NEWS / SCIENCE & SPACE

See also: science & space, antibiotic resistant bacteria

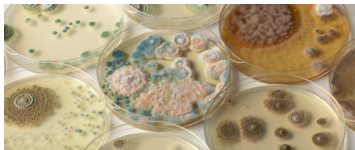
Researchers find a way to reverse antibiotic resistance in bacteria

May 6, 2015

Scientists May Have Just Found A Solution For Deadly Superbugs

The Huffington Post | By [Anna Almendrala](#) (@annaalmeidrala) ([user on Twitter](#))

Posted: 05/07/2015 5:06 pm EDT | Updated: 05/07/2015 6:59 pm EDT



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Biology leads to New Mathematics

Time Machine Problem: Given square matrices M_1, \dots, M_n , maximize a particular matrix entry over all n^ℓ words $M_{i_1} M_{i_2} \cdots M_{i_\ell}$ of length ℓ in these matrices.

What is the **computational complexity** of this problem ?

Does there exist an efficient dynamic programming approach?

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The answer is **no**. It can be reduced to k -SAT.

Theorem (Ngoc Tran & Jed Yang, arXiv: May 11, 2015)
“Antibiotics Time Machine Problem is NP-hard”.

Tran and Yang also show that a certain natural **convex relaxation** can be solved in **polynomial time**.

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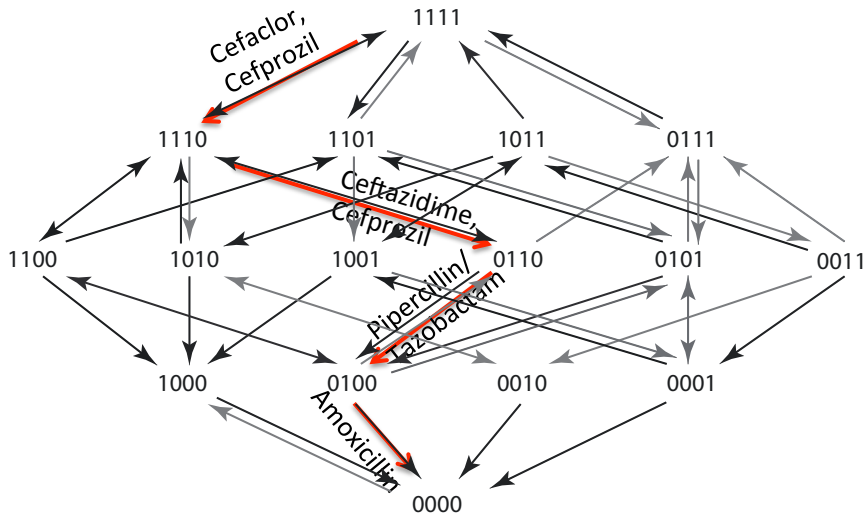
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¹Is Theoretical Computer Science Part of Mathematics?

Conclusion: Optimal Treatment Plans Can Be Designed



.... by solving the Time Machine Problem



15/05/2015

Resistência de bactérias aos antibióticos é revertida

Redação do Diário da Saúde

Evolução às avessas

A proliferação das [bactérias resistentes aos antibióticos](#) é um problema crescente em nível mundial.

A boa notícia é que esse problema talvez seja reversível.

Miriam Barlow (Universidade da Califórnia) e Kristina Crona (Universidade Americana) descobriram uma maneira de fazer com que as bactérias retornem a um estado pré-resistente.

As duas pesquisadoras demonstraram essa "evolução às avessas" usando uma família de 15 antibióticos disponíveis para combater infecções comuns, incluindo a penicilina.

Ciclagem de antibióticos

Para anular o processo r
pesquisadoras as expusi
calcularam a probabilidade



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Mon

May 6, 2015

Reverse Resistance by Rewinding Evolution