Is Antibiotic Rotation Optimal?

A mathematical approach to a medic's question.

Robert Beardmore, Dept of Maths Imperial College, London

One idea

October 1997, Michael Niederman's editorial...

- The `crop rotation' theory of antibiotic use has suggested that if we routinely vary our `go to' antibiotic in the ICU, we can minimize the emergence of resistance ...
- selection pressure for bacteria to develop resistance to a specific antibiotic would be reduced as organisms become exposed to continually varying antimicrobials."

MICHAEL S. NEIDERMAN, M.D. Director, Medical and Respiratory ICU Winthrop-University Hospital Mineola, New York Professor of Medicine SUNY at Stony Brook Stony Brook, New York

Is "Crop Rotation" of Antibiotics the Solution to a "Resistant" Problem in the ICU? Am J Respir Crit Care Med Vol. 156. pp. 1029–1031, 1997

And rotation is ...?

2005, JAC, Brown & Nathwani

- We identified 11 articles in which the authors claimed to have evaluated the efficacy of this intervention."
- Only four were suitable for review, but, owing to multiple methodological flaws and a lack of standardization, these studies do not permit reliable conclusions ..."

Journal of Antimicrobial Chemotherapy (2005) 55, 6–9 doi:10.1093/jac/dkh482 Advance Access publication 5 November 2004



Antibiotic cycling or rotation: a systematic review of the evidence of efficacy

A theoretical study

October 1997, PNAS, Bonhoeffer, Lipschitz & Levin

When more than one antibiotic is employed, sequential use of different antibiotics in the population ('cycling') is always inferior to treatment strategies where, at any given time, equal fractions of the population receive different antibiotics."

> Proc. Natl. Acad. Sci. USA Vol. 94, pp. 12106–12111, October 1997 Medical Sciences

Evaluating treatment protocols to prevent antibiotic resistance

SEBASTIAN BONHOEFFER^{†‡§}, MARC LIPSITCH[‡], AND BRUCE R. LEVIN[‡]

Models

- We want to understand the 'best' way of scheduling the deployment of antibiotics into some unit of treatment (hospital, ICU, [patient])
- We need a state-space for this model, what to put in it? Apologies, Pentti, for the abstraction, but let's not worry for now.

Journal of Antimicrobial Chemotherapy (2005) 56, 257–258 doi:10.1093/jac/dki230 Advance Access publication 21 June 2005 JAC

Mathematical model-tell us the future!

Pentti Huovinen*

Therefore the first comment on Magee's model concerns a lack of accurate data on antimicrobial consumption. I miss most a definition of the bacterial species that this theory concerns. Readers want to see how the model applies to the real world. It is reasonable to assume that different bacteria will behave differently. The epidemiology of *Streptococcus pneumoniae* is different from that of *Escherichia coli* or *Staphylococcus aureus*. Most of us have *E. coli*

One of many model structures, formulate the question as a control affine, optimal control problem:

 $\dot{x} = f(x) + A(t) \cdot \mathcal{A}x + B(t) \cdot \mathcal{B}x$

Let A(t) be the % of A-treated patients, same for B(t), everyone is treated: A(t)+B(t) = 100%.

 $x=(x_1,x_2,...)$ will contain everything we think might be important, keep this abstract for now.

x contains everything we think important:

- # susceptible & infected hosts in hospital & community 0
- # infected or colonised by WT bacterium or SDR/MDR mutants 0
- # treated for infection i with drug j (some appropriate, some not) 0
- # treated with multi-drug combinations with drug pair (j,k)? 0
- drug-specific resistance mechanisms: de novo mutation or plasmid-0 borne?
- parameters: causes of pathogen transmission, clearance rates due to 0 treatment, disease-induced death rates, duration of empirical treatment, pathogen mutation rates
- explicit spatial description or use a mass action law? 0

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explicit spatial description or use a mass action law?

 $\dot{x} = f(x) + A(t) \cdot \mathcal{A}x + (1 - A(t)) \cdot \mathcal{B}x,$ = $f(x) + \mathcal{B}x + A(t) \cdot (\mathcal{A} - \mathcal{B})x,$

- Output of the must-treat constraint, there's only one control variable: A(t).
- We now minimise `something' wrt A(t):

$$\min\left\{\frac{1}{T}\int_0^T(\varphi, x(t))dt: 0 \le A \le 1\right\}$$

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observable: continuous linear functional

optimal control problem $\dot{x} = f(x) + \mathcal{B}x + A(t) \cdot (\mathcal{A} - \mathcal{B})x,$ $\min\left\{\frac{1}{T}\int_{0}^{T}(\varphi, x(t))dt : 0 \le A \le 1\right\}$

symmetries lead to A-independence

- rotation: A(t) = 0 or 1: everyone is treated with the same drug at the same time
- mixing: A(t) = constant between 0 and 1: at least two patients receive different treatment for same infection

mixing strategies are used in practise & trials have implemented `maximal drug heterogeneity' policies

random mixing (A = 1/2) has been stated as optimal when compared with strictly periodic rotation

Iet's relax the assumption of strict periodicity:

Journal of Antimicrobial Chemotherapy (2005) 55, 6–9 doi:10.1093/jac/dkh482 Advance Access publication 5 November 2004

JAC

Antibiotic cycling or rotation: a systematic review of the evidence of efficacy

Erwin M. Brown¹* and Dilip Nathwani²

(xi) Is the order of rotation critical?

(xii) What is the optimal duration of each cycle? If cycles are too short, changes in resistance rates may not be detected. For example, in the study carried out by Bradley *et al.*,¹¹ the reduction in the incidence of colonization with GRE was most marked in the latter half of the second phase of the cycling protocol. On the other hand, too long a duration may be associated with high risks of resistant strains emerging during the cycle.

(xiii) Should the duration of each cycle be the same? (Does resistance to different antibiotics develop at the same rate?)

So As bang-bang (on-off) functions are weak* dense in the interval [0,1] within $L^{\infty}(0,T)$ and $\Phi(x(\cdot))$ is continuous wrt weak* topology, there are infimising sequences that rotate antibiotics.

The (smooth) equation

 $\dot{x} = f(x) + \mathcal{B}x + A(t) \cdot (\mathcal{A} - \mathcal{B})x, \qquad x(0) = x_0,$ defines an operator $x: L^{\infty} \to W^{1,\infty}; A \mapsto x(A)$

The functional we want to minimise, then, is

$$L^{\infty} \to \mathbb{R}; A \mapsto \Phi(x(A)$$
 where $\Phi(x) = \frac{1}{T} \int_0^T (\varphi, x(A)(t)) dt$

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homogensiation limit

a surprising consequence ? ...

Theorem If Φ is the continuous linear functional that denotes the performance measure, then

$$\inf_{A \in \{\text{periodic cyclings}\}} \Phi(x(A)) \le \inf_{A \in \{\text{mixings}\}} \Phi(x(A))$$

Proof. The weak* closure of the set of periodic, bang-bang functions contains all the constant functions. \Box

note the inf, it's not a min



Ecological theory suggests that antimicrobial cycling will not reduce antimicrobial resistance in hospitals

Carl T. Bergstrom*¹, Monique Lo[‡], and Marc Lipsitch[‡]

*Department of Biology, University of Washington, Seattle, WA 98195; and *Harvard School of Public Health, Boston, MA 02115

 $\dot{x} = f(x) + \mathcal{B}x + \omega \cdot (\mathcal{A} - \mathcal{B})x,$ $-\dot{\mu} = \varphi + \nabla f(x)^T \mu + \mathcal{B}^T \mu + \omega \cdot (\mathcal{A}^T - \mathcal{B}^T)\mu,$ $x(0) = x_0, \mu(T) = 0$

The EL equations could have a totally singular mixing control, w, but under what conditions?

there are trivial symmetries, any others?

define the Hamiltonian

 $\overline{H(x,\mu,A)} = (x,\varphi) + (\mu,f(x) + \mathcal{B}x) + A(\mu,(\mathcal{A} - \mathcal{B})x)$

Pontryagin's Maximum Principle states that

 $H(x^{*}, \mu^{*}, A^{*}) = \max_{0 \le A \le 1} \overline{H(x^{*}, \mu^{*}, A)}$

and so $A^{*}(t) = 1$ or $A^{*}(t) = 0$ unless

 $(\mu,(\mathcal{A}-\mathcal{B})x)\equiv 0$ & so $(arphi,(\mathcal{A}-\mathcal{B})x(0))=0$

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 $(\mu, (\mathcal{A} - \mathcal{B})x) \equiv 0$) & so $(\varphi, (\mathcal{A} - \mathcal{B})x(0)) = 0$

DAEs

P-IC symmetries

Let's apply these symmetries to a model taken from Bergstrom, Lo, Lipschitz; PNAS 2004:

 $\dot{S} = \mu(m-S) - (\tau_1 + \tau_2 + \gamma)S + \beta SX + \sigma\beta(c_1R_1 + c_2R_2)S,$ $\dot{R}_1 = \mu(m_1 - R_1) - (\tau_2 + \gamma)R_1 + \beta(1 - c_1)R_1X - \sigma\beta(c_1S + (c_1 - c_2)R_2)R_1,$ $\dot{R}_2 = \mu(m_2 - R_2) - (\tau_1 + \gamma)R_2 + \beta(1 - c_2)R_2X - \sigma\beta(c_2S + (c_2 - c_1)R_1)R_2,$ $\dot{X} = \mu(1 - m - m_1 - m_2 - X) + (\tau_1 + \tau_2 + \gamma)S + (\tau_2 + \gamma)R_1 + (\tau_1 + \gamma)R_2 - \dots$ $\dots - \beta X(S + (1 - c_1)R_1 + (1 - c_2)R_2),$

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X uncolonised, R_1 and R_2 are infected by drug 1 and drug 2 resistant bacteria, S by drug-susceptible bacteria.

Table 1: description of model parameters

parameter	meaning
$ au_1, au_2$	rate of use of drugs 1 and 2 per unit time (days)
m, m_1, m_2	patients enter hospital in states S, R_1 and R_2
	at rates $\mu m, \mu m_1$ and μm_2 resp.
c_1, c_2	fitness cost of resistance to bacteria
σ	relative rate of secondary colonization to primary colonization
eta	rate constant for colonization of uncolonized individuals
γ	untreated patients colonized by susceptible bacteria remain
	colonized $1/\gamma$ days on average
μ	rate of patient turnover in the hospital
α	represents physician compliance with cycling program

The must-treat constraint means $\tau_1 + \tau_2 = \tau_{max}$

and we want to minimise $\int_0^T R_1(t) + R_2(t) dt$

Mixing is at least codimension-3

Theorem

Suppose that parameters and initial conditions $(S(0), R_1(0), R_2(0), X(0))$ are nonnegative and suppose there is a mixing optimal control τ_1^* that we denote by the constant $\omega^* \in (0, \tau_{\max})$. Then

$$R_1(0) = R_2(0), \ c_2 = c_1 + \frac{\tau_{\max} - 2\omega^*}{\beta(X(0) + \sigma S(0))}, \ m_1 = m_2 + \frac{2(2\omega^* - \tau_{\max})\sigma R_1(0)^2}{\mu(X(0) + \sigma S(0))}.$$

As a result, if the 50-50 mixing protocol $\omega^* = \frac{\tau_{\text{max}}}{2}$ is optimal then $c_1 = c_2$ and $m_1 = m_2$, therefore $R_1(t) = R_2(t)$ for all $t \ge 0$.

a 'universal' result

antibiotic rotations are as good as any set of strategies

If or asymmetric models, there is a weak* nbd of ∞many rotations all better than mixing:



a rule-based controler

for asymmetric models, suboptimal rotations can outperform optimal mixing:





rule: if $R_1 > R_2$, use only drug 2

the controller beats mixing if information quality is high enough





So what?

- In a sense Niederman was right, but how to determine the best rotations in practise?
- Is this kind of result even relevant to current medical thinking?
- What model structures undo the veracity of this conclusion?

We developed a spatially-explicit, stochastic model to test experimental/computational proxies of protocols from recent clinical trials: based on a gradostat protocol.

protocols

- (0.) *null protocol*: no patient is treated;
- (1.) random sequential treatment: each patient receives a random drug each day (mixing);
- (2.) *empirical treatment*: a random drug is allocated to the patient the moment the patient arrives at the hospital (mixing);
- (3.) *periodic cycling*: scheduled, rotating cycles of antibiotic prioritisation and restriction are fixed before any patients have entered the hospital (rotation);
- (4.) periodic antibiotic monitoring and supervision: the next patient to arrive at the hospital will be treated with the drug that maximises the heterogeneity of drugs currently used within the hospital (mixing);
- (5.) *surveillance-based rotation*: drug-resistance phenotypic assays are regularly conducted on samples from the patient cohort and the same drug, namely the one estimated to have the currently lowest prevalence of resistance, is prescribed to all patients (rotation);
- (6.) DNA diagnosis-based dynamic treatments: an initial and rapid assessment of the bacterial genotypes responsible for each patient's infection is made the moment a patient arrives at the hospital, a reassessment may also be conducted later during the treatment (mixing).

spatially-explicit model, synergistic drugs



therapy given until pathogens below threshold : _____



: each of 100 'states' (the queue) is infected by SDR and MDR pathogens



: a new state enters the simulation

spatially-explicit model, synergistic drugs

random empirical therapy is better than none:







the best treatment (for this queue) gives the most appropriate therapy ASAP: range of cycling





single drug resistance

some rotations are good, many are not:





: domain size effects







shortest MLoS

v of resistance

123456

bang-bang controls: antibiotic mixing

many thanks

Rafael Pena-Miller, Conacyt-funded
Sylvain and Florence, CNRS Montpellier
Craig Maclean, Oxford
Martin Ackermann, ETHZ
MRC, EPSRC, NSF - Nescent

weak* convergence

We say that $(\phi_n) \subset L^{\infty}$ converges weak* to $\phi \in L^{\infty}$ if

$$\int_0^T \phi_n(t)\psi(t)dt \to \int_0^T \phi(t)\psi(t)dt \qquad \forall \psi \in L^1.$$

This is useful because if $\|\phi_n\|_{\infty}$ is bounded, (ϕ_n) has a weak* convergent subsequence.