MATHEMATICAL BIOLOGY

The subject is not as hierarchical as many areas of mathematics and there is some flexibly over what is included. It is an interdisciplinary sience

Classical areas:

- · population biology
- · epidemiolopy
- of several species · population interactions
- · population genetics (a) mathematical physiology Co] DNA modelling

Mathematical Back ground

- · liveor algebra
- · ODE & PPE
- o différence equations [0] vertor coloulus
- 6) probabilities & staticities.

Bismathematics as a methematics discipline

- · comparatevely new last 20 l 30 years.
- · increased number of advertised positions in academia, medicine and industry around the world
 - · new maners during as subject becomes so large, such as · Coofleried mechanics
 - · theoretical ecology
 - · Concer models growth models.



- · The mathematical descriptions of brological phenomeno are not biological explanations
- The principal rese of any theory is its predictions, this is why methematical models are important. However only experiments can confirm how dosely the model relates to the real biology.

| moderstanding | > mathematical | > predictions

The model can not take into account every single process - it captures the essence of the process that dominates the interations with more data avalable, more rephisticated models are possible. From mathematical point of view the act of good modelly relies on

(i) sound understanding and appreciation of the Ciological problem

(ii) a realistic mathematical representation of the important biological phenomeno

(iii) finding useful solutions, preferably quantitative (a) biological interpretation of the mathematical results in terms of insights & predictions.

The mathematics is dictated by the biology and not vice-versa.

- · Dynamic models are simplified representations of some real-world entity, in equations or computer and computer code
 - They minic some essential features of the study system while leving out inessentials.
 - . The models are colled dynamic becoure they describe how system properties change over time; e-g.
 - gene expression level

 - abundance of an endangered species mercury level in different organs within an individual ...
 - Allow for making forecasts that can not be made strictly by extrapolating from data

The modeling process:

- 1. Conceptual model represents our ideas about how the system works
- 2. Diagram with boxes or arrows
- 3. Equations are developed for the retes of each process and are combined to form a mathematical model consisting of dynamic equations for each state variable.

Compartment recodels- that based on conservation of mass: Example: Enryne Kinetis Enzyme mediated biochemical $S+E \stackrel{K_1}{=} SE$ SE P+E S = Substrate (sucrose) CpH20, E = enzyme (invertase) P = reaction product

CottpOb

(1) (Hydrolysis of sucrose Juto glucon & frictore) Only assumption: Law of Mars' Action - the rate of chemical reaction is proportional to the product of concentration of reactants K, K-, - comfacts of proportionality s, e, c, p -> concentrations of S, E, SE, P . Sand E combine to form SE at rate [K, Se] · SE separak & StE at rate SE — 11 — S+P — 11 we have 4 state variables 3, e, c, p rate of change = inflow rate - outflow rate E Lkies Ik,es (K,+K2) C

We can write now equations
$$\frac{dS}{dt} = K_{-1}C - K_{1}eS$$

$$\frac{de}{dt} = (K_1 + K_2)C - K_1es$$

$$\frac{dc}{dt} = \kappa_1 es - (\kappa_1 + \kappa_2)c$$

Nonlinear system of

Initial data:

$$c(o) = p(o) = 0$$

- · Our interest is in the rate of product formation, we do not need theegn. for p the rate of product formation is $k_2 c(t)$
 - · We con further simplify by nothing that $\frac{de}{dt} + \frac{dc}{dt} = 0 = 0$ ((t) + e(t) = ((0) + e(0) = e_0 =7 (elt) = e0 - C(t)

This makes sence beinge e and C are the unbound and bound forms of the enzyme which is not created or destroyed in this reaction scheme whereas substrate is irreversibly converted to product. We can substitute to-c for e into the equations for s and c leaving us with a system of only two equations

| as = K, (eo-c)s

 $\frac{dc}{dt} = \kappa(e_0 - c)s - (k_-, + k_2)c$

we can not solve these (except numerically) using computer) but we can get the iformation that we want by a simple approximation

Enzymes are in a very low concentration,

lo is small $e(t) = e_0 - c(t) \ge 0 \implies c(t) \le e_0$ $\Rightarrow c(t) \text{ is also small}$

We introduce neu variable

 $v = \frac{c}{e_0}$ => $c = e_0 V$

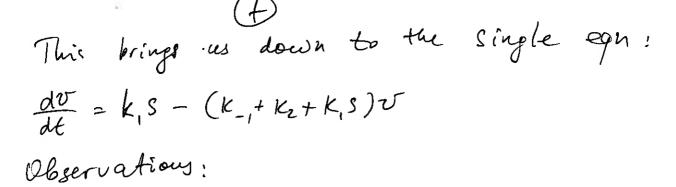
 $\frac{dv}{dt} = \frac{1}{e_0} \frac{de}{dt} = K_1 \left(\frac{e_0}{e_0} - \frac{c}{e_0} \right) S - \left(K_1 + K_2 \right) \frac{c}{e_0}$

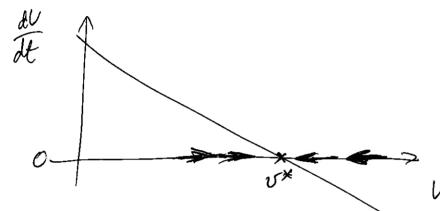
 $\frac{dv}{dt} = k_1 \left(1 - v\right) S - \left(k_1 + k_2\right) v$

The other equation becomes

15 = 15, eou - K, (eo-eou)s = eo[k_v-5,(i-v)s]

=> 5 changes much more slowly because eo is small => we can proceed as if s were compant (5 × court.)





oft) approaches value or at which de = 0

K,S = (K,S + K, + K2) 5 = (K,S+K-,+K2) = (K,S+K-,+K

$$C = \frac{K_1 e_0 S}{K_1 + K_2} = \frac{e_0 S}{K + S}$$
, $K = \frac{K_1 + K_2}{K_1}$

c* is collect stable equilibrium point: the system tends toward it and then stays there.

We need rate of product formation

$$\left(\frac{dP}{dt}\right)_{k} = k_{2}C^{*} = \frac{k_{2}los}{K+s} = \frac{V_{max}s}{K+s}$$
 where $\left[\frac{k_{2}los}{k_{2}los}\right]_{max}$

Michaelis - Menten equation.

We have some additional information:

the c-equation describes bow enzyme

moves between bound and renbound states—

it reaches equilibrium value, that depends

on S (the ammount of substrate)

Physics Models

Based on the physical laws of Mechanics. The variables are normally positions and velocities, accelerations, forces, etc. Equations come from Newton's laws of motion.

Example: DNA molecule

Simple model: A chain of coupled disles

MI N-1

MIN N-1

MIN N-1

MIN N-1

Un and In ave longitudinal and augulor displacements of the n-th disk

M and I are the mass and the moment of inertia of the disk

Le clasticity constant of spring

K-torgional rigidity of the chain 9

Mun = K (un+, -2un + un-) linear approx.

Force acting on n-th disk dutto

acceleration (clasticity)

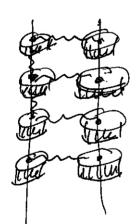
of nth node

This is a linear approximation

 $I \varphi_n = K \left(\varphi_{n+1} - 2 \varphi_n + \varphi_{n-1} \right)$ torque

augular acceleration

These two equations give the linear approximation of one of the DNA strands. But DNA has two strands, which have a planar representation—in the next approximation:



Each Disk represents a base of DNA

Actually the real DNA molecule is by for more complex. It is a doloble the helix with a turn of about 36° per base.

The Watson - Crick Model of DNA 1953 DNA composed of 4 bases (or nucleotides)

AT, C G which pair according to the A-T C-G two complementary strands $C = A = T = G = T = C = \frac{3}{2}$ G = T = A = G = AStrand 1: Strand 21 3.4×100 m very strong sugar-phosphate backbones

Each cell contains 10" base pairs = 3.4 m of DNA in each cell of over body! Actually it is between 1-2 meters and not in one piece.

The total volume however of DNA is The total volume however of DNA is the total value and can be packed still rather small and can be packed in individual cells with plenty of space left

One gene encodes one entire protein and a typical lenth of a gene is 500-600 bp (basic pairs) One bp > 60 atoms.

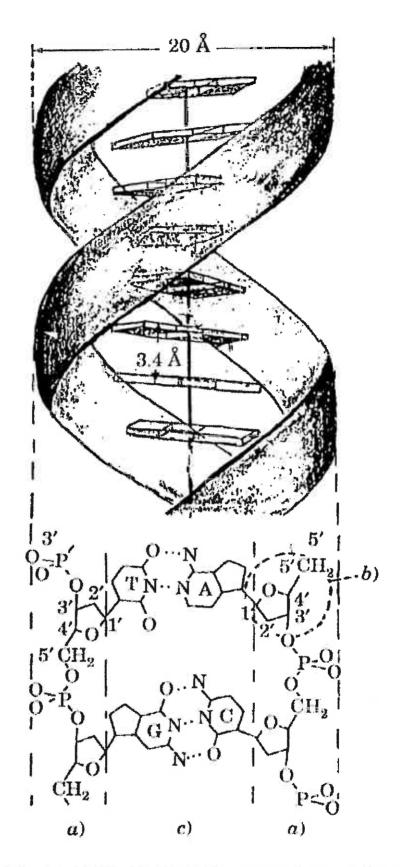
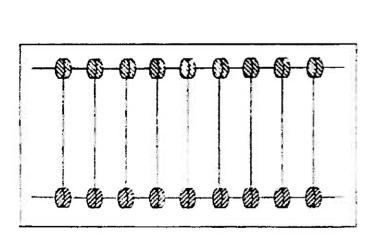
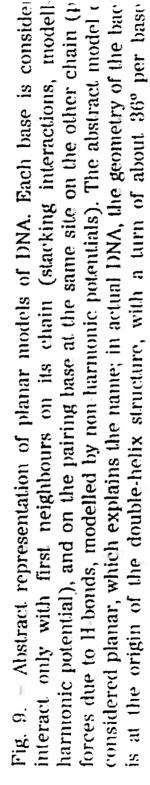


Fig. 1. – Schematic drawing of the double helix (B form), the sugar-phosphate backle represented as a ribbon, with detailed chemical structure shown in a) and b) (sugar notice bases are sketched as rectangular plates, with detailed chemical structures shown







pairing interactions indicated in fig. 9, these consider also interaction between bas nearby in three space due to helical geometry; two bases connected by such an shelic Interaction are indicated. In real DNA, these interaction are mediated by water fila: Fig. 10. — Abstract representation of helicoidal models of DNA. In addition to stackin opposite chains which are half pitch of the helix away in the sequence, and the oining the concerned bases. In our mechanical model: # DNA 1'S smoothed or averaged to year a model as an classic line.

. DNA is big enough to be seen with various microscopy techniques

DNA quite offen occurs into closed loops (150 bp or less), mini-circles as well as thousands of bp

. DNA is rasker stiff; it has a well-defined share shape.

Optimization Models

Biologists also use optimization models in which the modeler assumes that the Study organizus are trying to achieve some particular goal

· For example; on animal will gather food in the babitat that gives it the Best chance of getting enough pood to survive, without itself berng eaten by predators.
Optimization of drug usage: effect/damage.
Usage of dynamical modelds

Why do we need dynamical models? 1. Scientific renderstanding of the underlying process or phenomenon

2. Using scientific understanding to manage the world

· Forecosting disease or pest outbreaks · Designing man-ruade systems, for example biological pest control, bioingeneering

· Managing existry systems such as africulture or fisheries

· Optimitig medical treatments or improvize athletic performances

3. There are experiments that you would rether not do:

· Endangered species management by trial &

· Clinical experiments on bumans

experiments with toxic substances, radiological experiments, losymetry and vadiation protection.

Theoretical versus

Practical Models

Practical Main goals are management, design & prediction

- · Numerical acuracy is desirable even at the expense of simplicity
- · Processes and details can be ignored only if they are numerially unimportant
 - of system processes
 - · System and question specific

Theoretical

· Main pools are theoretical understanding and theory development

- essential; the model should be as simple as possible.
 - · Processes and details con be ignored if they are conceptually irrelevant to the theoretical issues
 - Assumptions may be qualitative representations of hypotheses about the system, adopted conditionally in order to work out their consequences
 - · Applies to a range of Similar systems

Money

Out line of the modellity process > identify your objective Fis it fearable? Construct an initial rudel El aluate the model Accept as is Reject & start again