

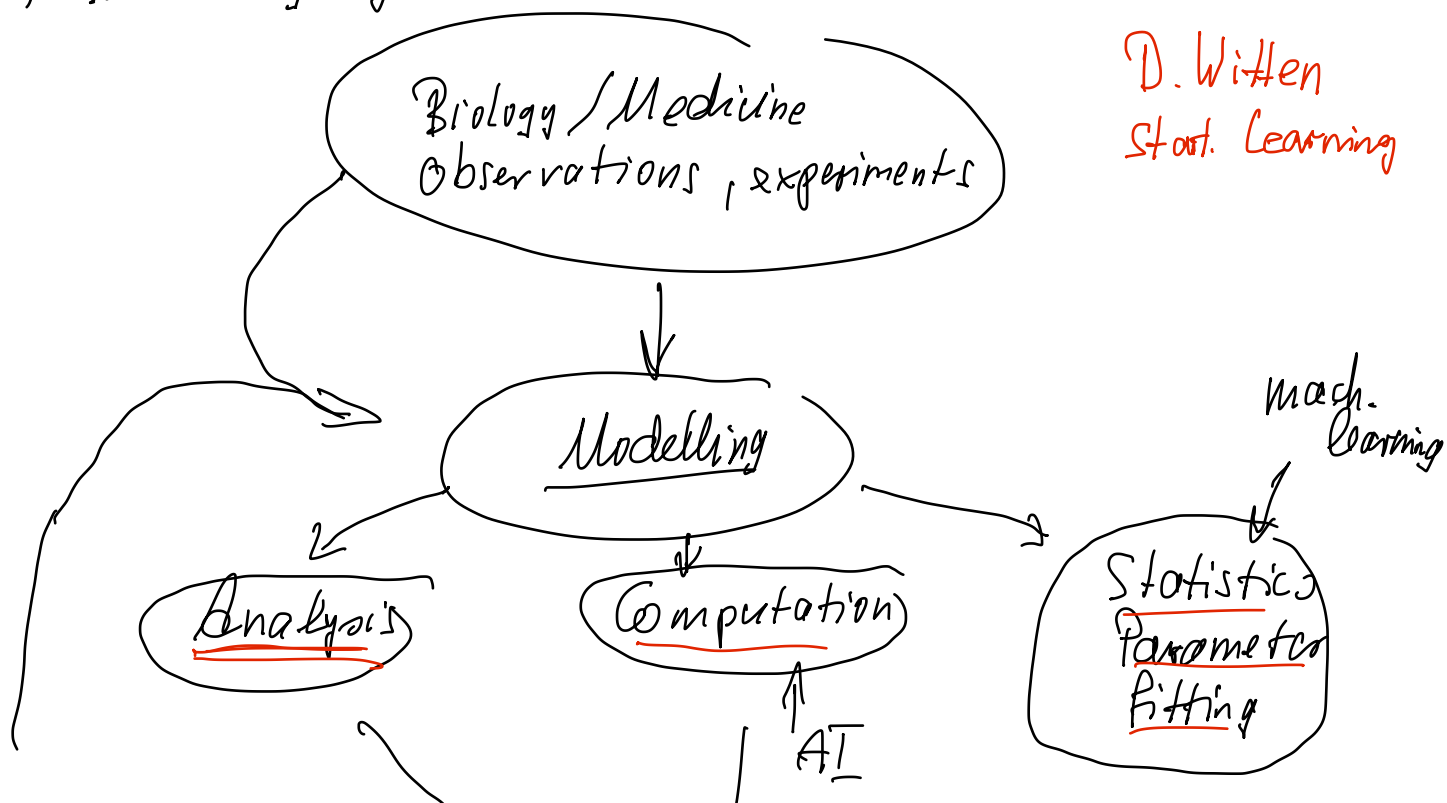
Math Oncology

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Starting SOON.

of course today: Computer problems.
Coming soon.

- 1) Introduction.
- (1.1) Course outline
- (1.2) Modelling cycle

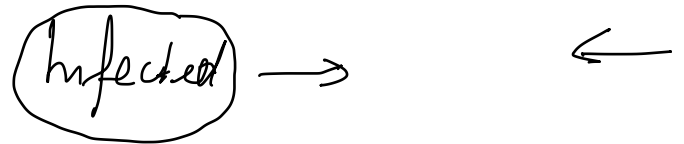


Insight, Explanations
Control

(1.3) Probabilities and Rates



quantify transitions:
recovery



- assume 2 out of 100 people recover per day.

$$\text{prob. of recovery } p_1 = \frac{2}{100} = \frac{1}{50}$$

$$\alpha_1 = \text{rate} = \frac{\text{probability}}{\text{unit of time}} = \frac{1}{50} \frac{1}{\text{day}}$$

- 2 days. prob. $p_2 = \frac{4}{100} = \frac{1}{25}$

$$\alpha_2 = \text{rate} = \frac{\text{prob}}{\text{unit of time}} = \frac{1}{25} \frac{1}{2 \text{ days}} = \frac{1}{50} \frac{1}{\text{day}}$$

- $\frac{1}{2}$ day $p_{1/2} = \frac{1}{100}$

$$\alpha_{1/2} = \frac{1/100}{1/2 \text{ day}} = \frac{1}{50} \frac{1}{\text{day}}$$

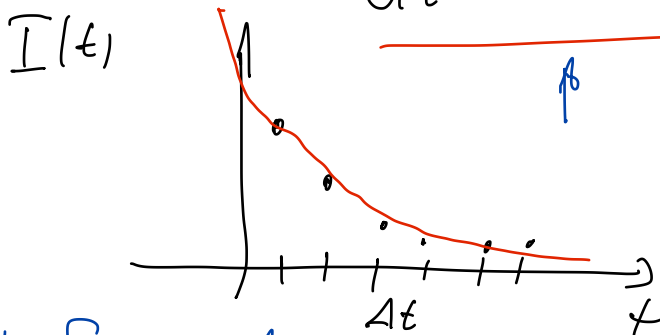
generalize Δt , $P_{\Delta t}$, $\alpha_{\Delta t} = \frac{P_{\Delta t}}{\Delta t} = \alpha$ (*)

Master equation: $I(t + \Delta t) = I(t) - P_{\Delta t} I(t)$

$$\frac{I(t + \Delta t) - I(t)}{\Delta t} = - \underbrace{\frac{P_{\Delta t}}{\Delta t}}_{\alpha} I(t)$$

$$\frac{dI}{dt} = - \alpha I$$

$$I(t) = I_0 e^{-\alpha t}$$



(1.4) Survival - Time analysis

(Sojuorn time, Thieme book)

a : time an individual spends in a state

$F(a)$: prob. to still be in the state at time a .

$F(a)$ non-increasing $F(0) = 1$

$$\lim_{a \rightarrow \infty} F(a) = 0 \quad \leftarrow$$

If $F(a) = 0 \quad \forall a > c$, then c is a maximum duration in this state.

T : random time of stage exit.

$$F(a) = \Pr(T > a)$$

$$G(a) = 1 - F(a) \text{ prob. of exit} \\ = \Pr(T \leq a)$$

Cond. probability to survive h time units longer, if survived already a time units is

$$F(h|a) = \frac{F(a+h)}{F(a)}$$

prob. exit between a and $a+h$

$$1 - F(h|a) = \frac{F(a) - F(a+h)}{F(a)}$$

define exit rate

$$\mu(a) = \lim_{h \rightarrow 0} \frac{1}{h} (1 - F(h|a)) \\ = \lim_{h \rightarrow 0} \frac{F(a) - F(a+h)}{h F(a)}$$

$$\mu(a) = - \frac{F'(a)}{F(a)}$$

Example 1: (exponential distribution) Assume the survival time is exponentially distributed.

$$F(a) = e^{-\gamma a} \quad \gamma > 0$$

$$\mu(a) = \frac{\gamma e^{-\gamma a}}{e^{-\gamma a}} = \gamma \text{ exit rate}$$

Conditional prob. $F(h|\alpha) = e^{-\gamma h} = F(h)$

Theorem (Thieme book)

→ $F(h|\alpha)$ is independent of α iff $F(\alpha) = e^{-\gamma \alpha}$
for some $\gamma \geq 0$.

If Δt is small we can write the prob. of leaving

$$Q(\Delta t) = 1 - F(\Delta t|t) = 1 - e^{-\gamma \Delta t}$$

more correct

$$\approx 1 - (1 - \gamma \Delta t + \text{h.o.t.})$$
$$= \gamma \Delta t + o(\Delta t)$$

⇒ $\boxed{\gamma = \frac{Q(\Delta t)}{\Delta t}}$ this is (*)

approx for Δt small

Back to epidemic model

$$I(t + \Delta t) = I(t) - Q(\Delta t) I(t)$$

$$\frac{I(t + \Delta t) - I(t)}{\Delta t} = - \frac{Q(\Delta t)}{\Delta t} I(t)$$

$$\dot{I} = -\gamma I$$

Important

1) A relation of the form

$$\dot{x} = -\gamma x$$

assumes exponentially dist. survival times.

2) The prob. to leave a state in the time interval $[t, t + \Delta t]$ is often denoted as $p_{\Delta t} = \gamma \Delta t$ ←

This is an approximation of

→
$$P_{\Delta t} = 1 - e^{-\gamma \Delta t} \quad \text{for small } \Delta t.$$

(1.5) Cancer

[→ Kulesza, et. al. *Jo. Pharmacokinetics and Pharmacodynamics*, 2024.]

Thoughts: — heterogeneous pop. of dynamic interactions.
— microenvironment (stroma, immune cells, blood supply, ECM, extracellular matrix)

Cancer Modelling

- multiple scales: genes, proteins, cell, tissue, body scale
- top-down: data inform models
- bottom-up: processes inform models
- need to strike a balance between complexity, data availability, and interpretability.