

**COMS30127/COMSM2127**

**Computational Neuroscience**

## **Lecture 4: Differential equations**

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# Intended learning outcomes

- Understand differential equations as tools for modelling real-world systems.
- How to solve easy ordinary differential equations.
- Get an intuition for the typical dynamics in linear ODEs.

# Differential equation models of real-world systems

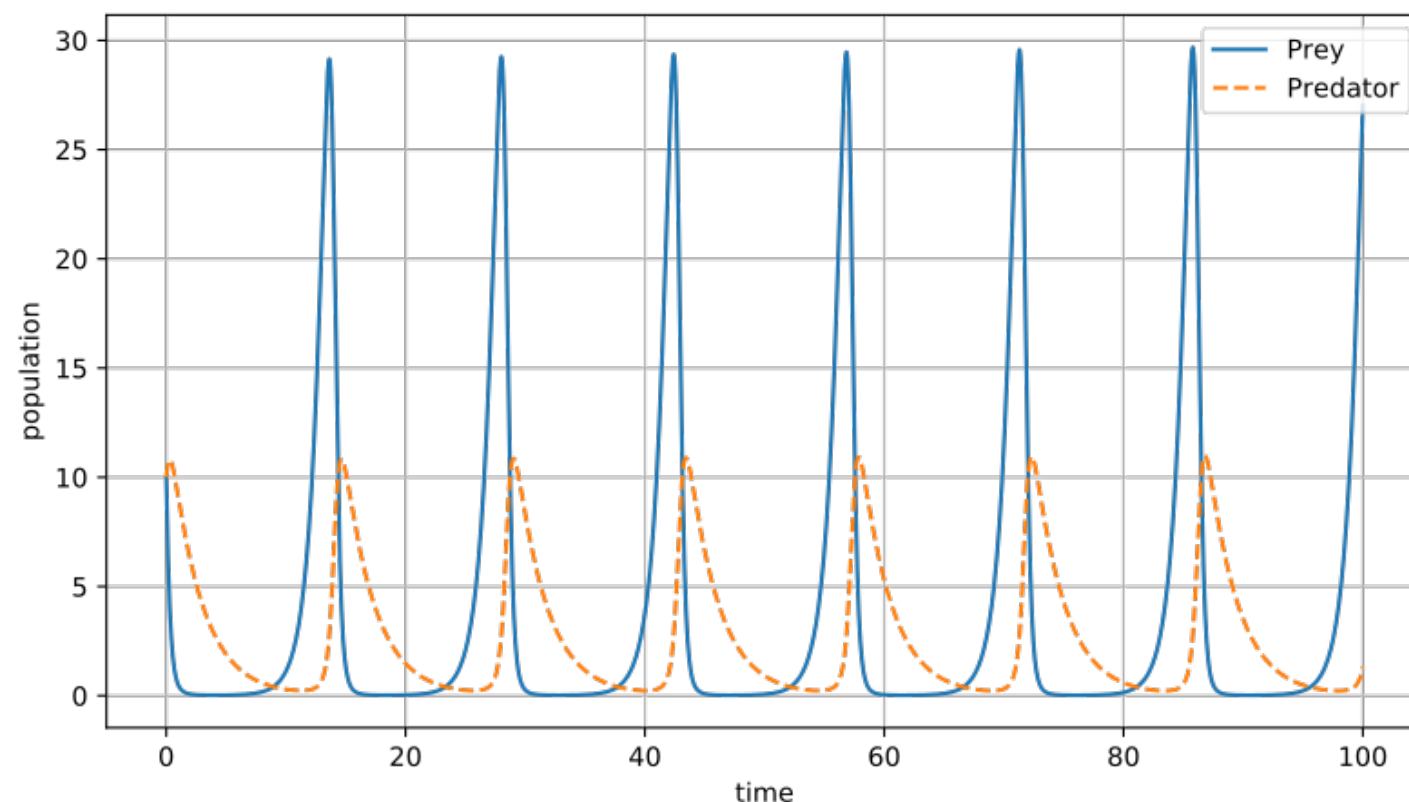
- Differential equations (DEs) are by far the most common formalism for modelling real-world systems.
- Ubiquitous in physics, chemistry, engineering, geoscience, biology.
- Useful for modelling things changing over time, and when the variables of interest can be treated as continuous quantities.
- However not universal! For discrete and/or stochastic systems we need other mathematical models.

# Lotka-Volterra model example

- Famous simple mathematical model of a predator-prey system from ecology.
- Consists of a system of two nonlinear ODEs:

$$\frac{dx}{dt} = \alpha x - \beta xy$$
$$\frac{dy}{dt} = \delta xy - \gamma y$$

- $x$  and  $y$  are variables that represent the number of rabbits and wolves, respectively.
- $\alpha, \beta, \gamma$  and  $\delta$  are parameters.



# Types of differential equations

- Common type we will study are ordinary, linear, first-order, inhomogeneous differential equations, which can often be written in form

$$\tau \frac{df}{dt} = -f(t) + g(t)$$

- *Ordinary* (vs partial) refers to the fact that there is only one dependent variable ( $f$  in the above).
- *Linear* (vs nonlinear) means there are no  $f^2$  or anything.
- *First-order* (vs second or third-order etc) means that the highest derivative is  $\frac{df}{dt}$ , there is no  $\frac{d^2f}{dt^2}$ , etc.
- *Inhomogeneous* because the solution depends on something else in addition to  $f$ . Usually in models of physical systems this will correspond to some external signal.

# Differential equation models in neuroscience

- In computational neuroscience the dynamic variables of interest are commonly things like:
  - Voltage and current flow in a single neuron.
  - Firing rates of a population of neurons.
  - EEG or fMRI signals (if modelling an experiment)
- For example here is a common differential equation for the voltage of a neuron:

$$\tau_m \frac{dV}{dt} = E_l - V(t) + R_m I(t)$$

# Solving easy differential equations

- “Solving” a differential equation means finding a function that when differentiated returns the original DE.

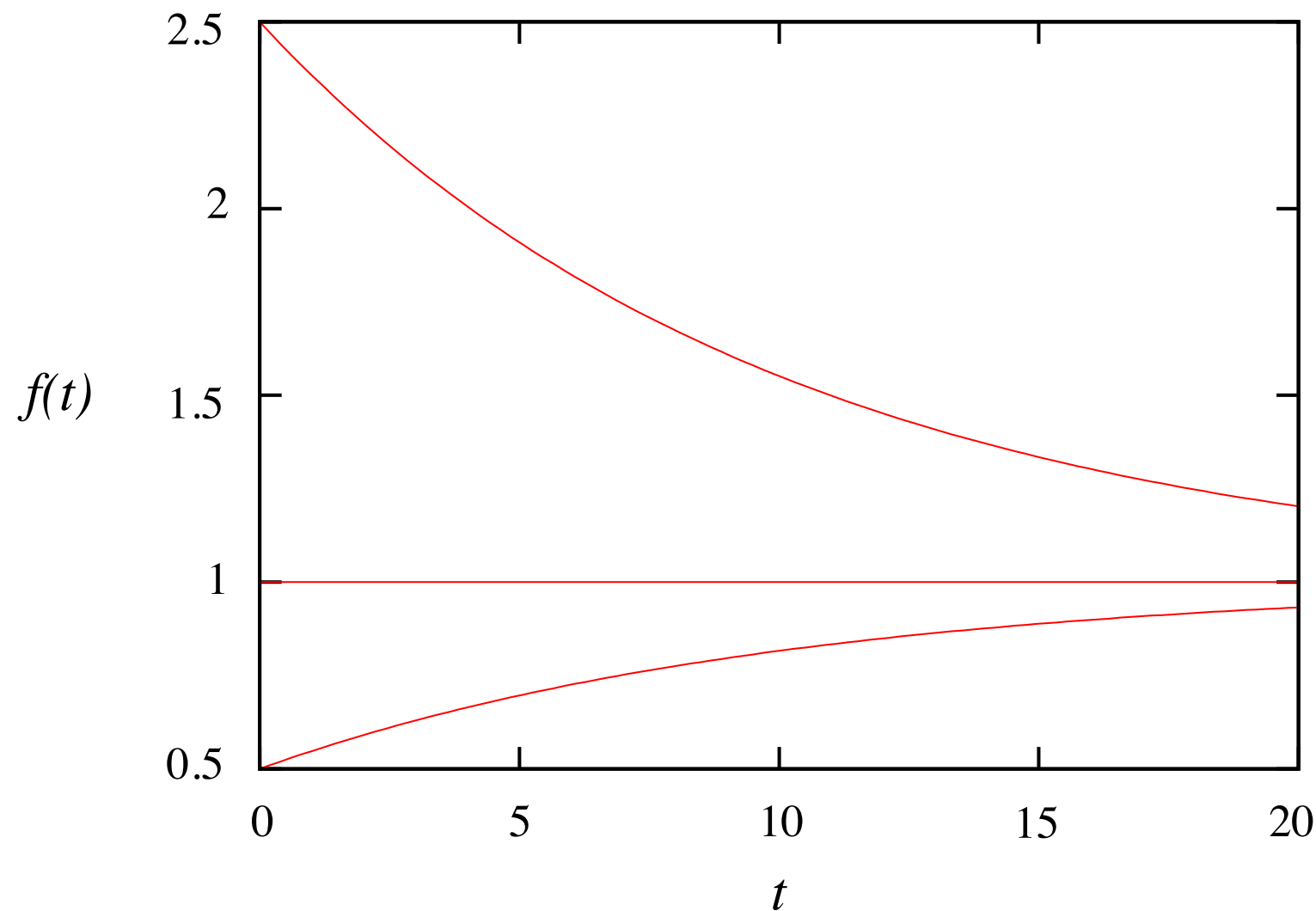
- In this case we want to find a formula for  $f(t)$ .

- So how do we solve  $\tau \frac{df}{dt} = -f(t) + g(t)$ ?

- See Conor Houghton’s notes on the GitHub repo:  
[https://github.com/coms30127/2019\\_20/notes/  
maths](https://github.com/coms30127/2019_20/notes/maths)

# The solution for constant $g$

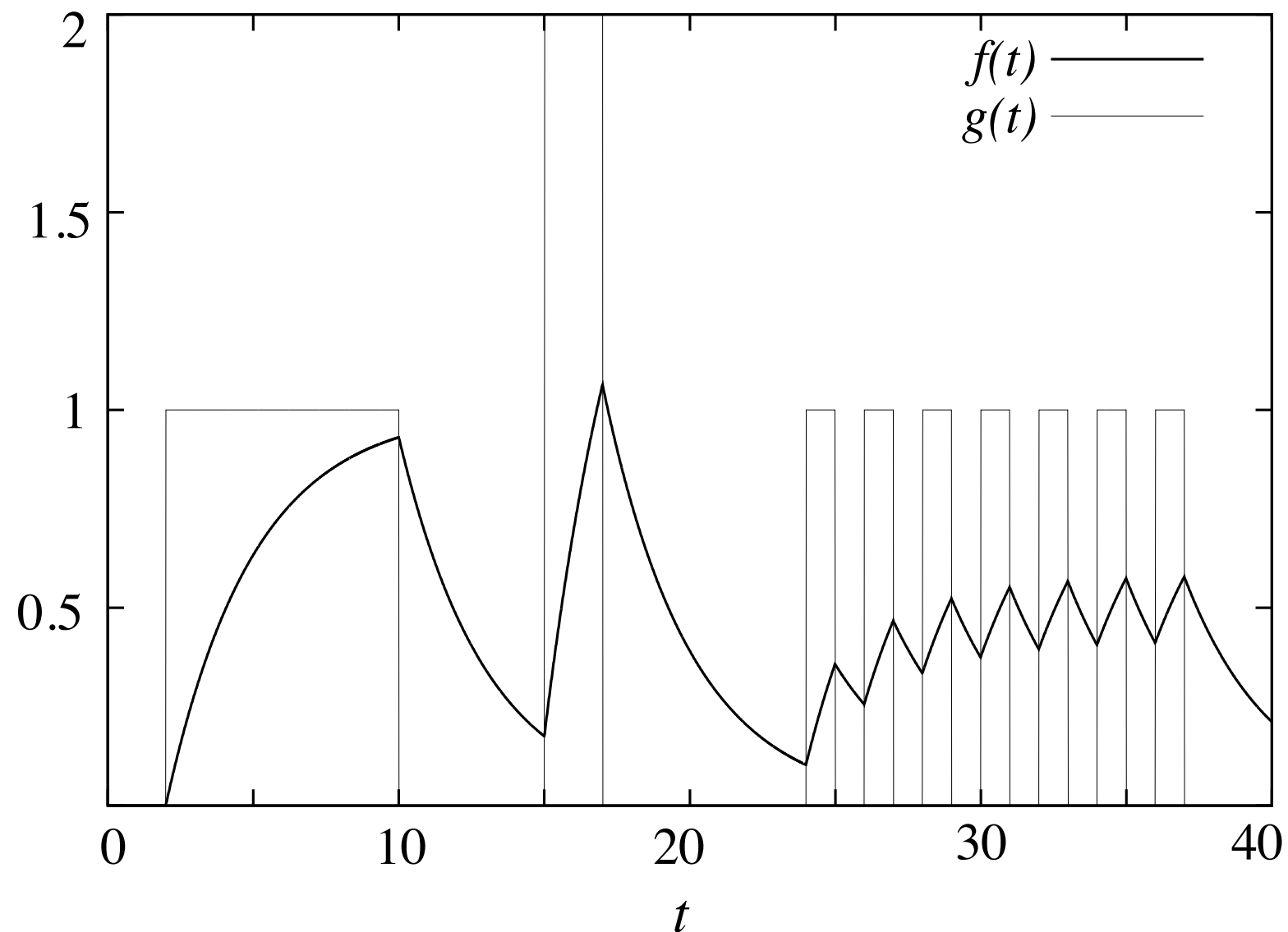
$$f(t) = \bar{g} + [f(0) - \bar{g}]e^{-t/\tau}$$



- Note that the trajectories all eventually converge towards the same value, no matter what the initial condition.



# The solution for time-varying $g$



- Again see Conor Houghton's notes on the GitHub repo: [https://github.com/coms30127/2019\\_20/notes/maths](https://github.com/coms30127/2019_20/notes/maths)
- Note that  $f(t)$  "tries" to track  $g(t)$ .

# Intuition for ODE dynamics

- Linear differential equations with a constant input do one of two things:
  1. Converge to a steady-state value.
  2. "Explode" to positive or negative  $\infty$ .
- Linear differential equations with a time varying input "try" to track the input signal.
- If the time constant(s) in the DEs are fast compared to the timescale of the input signal, the variables can track the input almost perfectly.
- In the converse case, if DE time constants are slow compared to the timescale of the input signal, then the variables will low-pass filter the input - they "average out" the high frequency components of the input.

# Next Tuesday

- In most models of practical interest, we can't solve the differential equations analytically.
- HOWEVER we can use computers to calculate an **approximate** answer numerically.
- Next Tuesday's lecture will introduce two basic numerical methods for solving ODEs, and mention some of the practical issues.