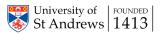
#### Applications of Mathematics

Theme Head: Irene Kyza

School of Mathematics and Statistics University of St Andrews



30 September 2025

#### Overview of the theme

The theme can be divided into two broad categories:

- the formulation and analysis of mathematical models
- methods or tools needed to analyse and simulate them

Two modules per category; self-contained and independent.

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Two major areas of modelling:

- continuum mechanics
- mathematical biology

Not comprehensive, but illustrative of a range of approaches.

Two complementary sets of methods:

- asymptotic and analytic
- numerical

Again not comprehensive, but illustrative. In both cases the aim is to develop *approximate* methods in a systematic, quantitative way.



## Continuum Mechanics (semester 1)

- O Continuum mechanics describes how deformable media behave:
  - Fluids (liquid/gas)
  - Solids (elastic/plastic)

Continuum hypothesis + Newton II:

- $\Rightarrow$   $O(10^{24})$  molecules: can't solve F=ma for each
- ⇒ Treat medium as a continuum of parcels, each small on scale of motion, each containing large number of molecules
- $\Rightarrow$  Then **u**, p,  $\rho$ , etc, considered as functions of  $(\mathbf{x}, t)$

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- Huge range of applications (nano-technology to astrophysics)

## Continuum Mechanics (semester 1)

#### Topics:

- Continuum mechanics: construction of dynamical models of deformable media
- Fluid Dynamics: lubrication theory, aerofoils, hydrodynamic stability
- Non-Newtonian fluids: fluid viscosity depends on internal stresses

#### Mathematical Biology and Physiology (semester 2)

- O Mathematical modelling in the life sciences.
- O Modelling of just about any aspect of biological systems:
  - Circulation (e.g., blood flow in arteries)
  - Patterns (e.g., rashes)
  - Populations dynamics
  - Cell dynamics (dynamics at cell scale)
  - Diseases/treatments (epidemiology)

#### Mathematical Biology and Physiology (semester 2)

#### Topics:

- Bacterial resistance: antibiotics
- Mathematical Physiology: microscale/macroscale & homogenisation
- Population Modelling: epidemiology, evolution, pathogen-host interactions, age-structured models
- Mathematical Oncology: cancer modelling

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Can change the character of the problem

$$i\epsilon\partial_t u + \epsilon^2 \Delta u + Vu = 0.$$

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Some examples:

 In PDEs (e.g. Navier–Stokes, Schrödinger), small terms may give rise to boundary layers or caustics.



• In forced oscillators, small forcings may lead to resonance.

These lectures will provide a toolkit for tackling such problems.



**EXAMPLE:** Asymptotic expansion of the error function:

$$\operatorname{erf}(x) = \frac{2}{\sqrt{\pi}} \int_0^x e^{-t^2} dt$$
 (< 1).

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- Convergent: gets arbitrarily close to erf(x) for any x
- x = 4: needs > 40 terms for a reasonable approximation
- x = 4: first 16 terms sum to  $10^5$



O Alternative approach: Write

$$\operatorname{erf}(x) = 1 - rac{2}{\sqrt{\pi}} \int_{x}^{\infty} e^{-t^2} dt$$

and integrate by parts *n* times:

$$\int_{x}^{\infty} e^{-t^{2}} dt = \int_{x}^{\infty} \frac{t e^{-t^{2}}}{t} dt$$

$$= e^{-x^{2}} \left[ \frac{1}{2x} - \frac{1}{4x^{3}} + \frac{3}{8x^{5}} + \dots + \frac{(-1)^{n-1}}{2^{n}} \frac{1 \cdot 3 \cdot 5 \dots (2n-3)}{x^{2n-1}} \right] + R_{n}$$

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• x = 4: first 2 terms give erf(4) to within  $10^{-7}$ 



#### Topics:

- Multiple scales: modulation and resonance
- Matched Asymptotics: boundary layers
- WKB Theory: rapidly oscillating solutions, ray theory
- Approximation of Integrals: Watson's lemma, steepest descent
- Intermediate Asymptotics: self-similarity of solutions
- Resummation: techniques for series

- O Ways to solve mathematical models numerically.
- O Often there is a *compromise* between ease of implementation and efficiency.
- O Or between speed (computational cost) and accuracy.

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#### Topics:

- Ordinary DEs: Explicit, implicit and multistep methods
- Stochastic DEs: Brownian Motion, Stochastic Integrals, Stochastic DEs and simulations
- Partial DEs: Finite Difference and Finite Element methods
- Linear Algebra: Numerical Solution of Linear Systems and Eigenvalue problems.

O Numerical methods for PDEs

Many models in applied mathematics are based on PDEs.

- Many pieces of software claim to be able to solve PDEs.
- Pretty pictures are not always accurate!!!
- It is good to know what are some standard things that can go wrong.

O Numerical methods for PDEs

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These lectures will look at some of the basic questions:

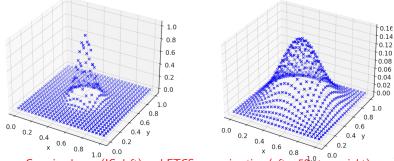
- How can we discretise PDEs to give systems of algebraic equations?
- How can we solve the resulting systems efficiently?
- How do we prove these methods have desirable properties, like
  - Consistency (representing the desired PDE)
  - Stability (not being hypersensitive to data)
  - Convergence (error decreasing as resolution increases)

#### Numerical Methods (semester 2): Finite Differences

O EXAMPLE: FTCS method for the heat equation:  $\partial_t u = \Delta u$  (homogeneous Dirichlet BCs):

$$\frac{U_{i,j}^{n+1}-U_{i,j}^n}{\Delta t}=\frac{U_{i+1,j}^n-2U_{i,j}^n+U_{i-1,j}^n}{(\Delta x)^2}+\frac{U_{i,j+1}^n-2U_{i,j}^n+U_{i,j-1}^n}{(\Delta y)^2}.$$

• If  $\Delta x = \Delta y$ , choose  $\Delta t < \frac{(\Delta x)^2}{2}$  to ensure *stability* and hence *convergence*.

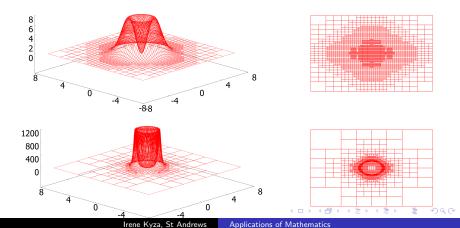


Gaussian bump (IC; left) and FTCS approximation (after 50 steps; right)

# Numerical Methods (semester 2): Finite Elements: More Advance Techniques!

#### Semilinear parabolic equations: Regional blowup

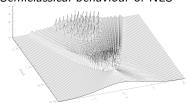
- $\partial_t u \Delta u = u^2$  in  $\Omega \times (0, T]$ ,  $\Omega = (-8, 8) \times (-8, 8)$
- $u_0(x,y) = 10(x^2 + y^2)e^{-0.5(x^2 + y^2)} \longrightarrow \text{Blowup set: circle centred on } (0,0)$



# Numerical Methods (semester 2): Finite Elements: More Advance Techniques!

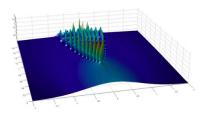
- Complicated geometries
- Sharply localised features

#### Semiclassical behaviour of NLS



Cai, McLaughlin & McLauglin, 2002

NLS dispersive breaking

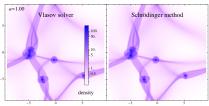


Bertola & Tovbis, 2013

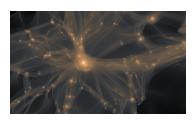
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- Complicated geometries
- Sharply localised features

#### Cosmology: Galaxy formation



Kopp, Vattis & Scordis, 2017



From the page of Dr. R. Kaehler

2d & 3d simulations for galaxy formation



#### Prior knowledge, delivery and assessment

Assumed knowledge of standard, core undergraduate material:

- Calculus, ODEs, PDEs
- Linear Algebra
- Complex Variables

More details on the module pages.

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Assessment: two assignments per module

- Mix of analytic and computational work
- Normally at least two weeks to complete each

