

Lecture 1: Introduction to the module

COMP5930M Scientific Computation

Today

Resources

Assessment

Scientific Computing

Nonlinearity

Newton's method

Module resources

- ▶ Staff
 - ▶ Dr Toni Lassila, t.lassila@leeds.ac.uk (personal queries)
- ▶ [Yammer](#)
 - ▶ Announcements
 - ▶ Questions about lectures and coursework answered
- ▶ [Minerva](#) (virtual learning environment)
 - ▶ Module information
 - ▶ Virtual lectures (pre-recorded)
 - ▶ Tutorials (live on Collaborate Ultra)
 - ▶ Coursework submission (through Turnitin)
 - ▶ Final assessment (through Gradescope)

Weekly learning pattern

Learning activities:

- ▶ Two video lectures (watch on Minerva)
 - ▶ Theoretical aspects of scientific computing
 - ▶ Watch on your own time, but suggested to reserve a fixed slot on your timetable every week to get into a routine
 - ▶ Questions about material? Ask and discuss in Yammer.
- ▶ Live tutorial session (Wednesdays at 11:05 UK time)
 - ▶ Practical examples of using MATLAB
 - ▶ Coursework hints and tips
 - ▶ Participation through Collaborate Ultra
 - ▶ Sessions recorded and can be reviewed later
- ▶ Weekly multiple choice quiz (Fridays on Gradescope)
 - ▶ Measure your learning every week
 - ▶ Does not factor into final grade (formative assessment)

Assessment

- ▶ Coursework 1 (20%)
 - ▶ Released: October 20th
 - ▶ Deadline: November 12th, 10:00 a.m. UK time
 - ▶ Submission electronically through Minerva
- ▶ Coursework 2 (20%)
 - ▶ Released: Nov 17th
 - ▶ Deadline: December 7th, 10:00 am UK time
 - ▶ Submission electronically through Minerva
- ▶ Final assessment (60%)
 - ▶ Online assessment (through Gradescope) in January
 - ▶ 48-hours to complete

What is scientific computing?

Use of computer algorithms to solve **mathematical problems** arising from science and engineering:

- ▶ **Computer algebra**
 - ▶ solution of linear and nonlinear system of equations
 - ▶ computational linear algebra
 - ▶ vector and tensor analysis
 - ▶ computational geometry
 - ▶ optimisation algorithms
 - ▶ linear/non-linear programming
- ▶ Numerical analysis
 - ▶ numerical solution of differential equations
 - ▶ numerical integration
 - ▶ discrete Fourier/Laplace transform

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The module syllabus

- ▶ Part 1: Solving nonlinear equations in one dimension
 - ▶ Solving one nonlinear equation: Newton's method
 - ▶ Alternatives: bisection method, secant method
 - ▶ Issues: convergence of Newton's method
- ▶ Part 2: Solving systems of nonlinear equations
 - ▶ Solving systems of equations: Gradient descent, Newton
 - ▶ Improvements: line-search, continuation methods
 - ▶ Partial differential equations: discretisation in space and time
 - ▶ Time-dependent problems: time-stepping algorithms

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- ▶ Part 3: Linear solvers and algorithms for large systems
 - ▶ Linear solvers: direct and indirect
 - ▶ Sparsity: computational complexity, pivoting, reordering
 - ▶ Direct methods: LU factorisation
 - ▶ Indirect methods: Gauss-Seidel, Jacobi, conjugate gradient

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Books

Not essential, all material contained in lecture notes.

Based on textbook:

Solving nonlinear equations with Newton's method,
CT. Kelley,
SIAM Fundamentals of Algorithms FA01, SIAM 2003.

There are a lot of *numerical methods/analysis* books which cover single nonlinear equations and some that briefly consider systems.

Numerical analysis,
RL. Burden and JD. Faires,
Brooks/Cole 2005 (8th edn).

Nonlinear equations

- ▶ What is nonlinearity?
- ▶ Why do we need numerical methods to solve nonlinear equations?

Nonlinear equations

- What is nonlinearity?

Prototype scalar equation: Find $x^* \in \mathbb{R}$ such that

$$F(x^*) = 0$$

for a given function f that is **differentiable** with derivative $F'(x) \neq \text{constant}$ (non-constant derivative)

Prototype system of equations: Find $\vec{x}^* \in \mathbb{R}^n$ such that

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with Jacobian matrix $\vec{F}'(\vec{x}) \neq \text{constant}$ (non-constant Jacobian)

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Examples of nonlinear equations

- ▶ Polynomial equations: $a_p x^p + a_{p-1} x^{p-1} + \dots + a_1 x + a_0 = 0$
(common approximations for other nonlinear equations)
- ▶ Trigonometric equations $\sin(x) + 2 \cos(x) = 0$ *etc.*

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- ▶ Many physics equations: $n_1 \sin \theta_1 = n_2 \sin \theta_2$, $E = mc^2$ *etc.*

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- ▶ Many physics equations: $n_1 \sin \theta_1 = n_2 \sin \theta_2$, $E = mc^2$ etc.
- ▶ Backpropagation for finding weights in a neural network:

$$\frac{\partial J}{\partial W} = \frac{1}{n} \sum_i^n x^{(i)} \left(W x^{(i)} + b - y^{(i)} \right) = 0,$$

where J is the loss function, $x^{(i)}$ the input and $y^{(i)}$ the output

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Closed form solutions of nonlinear equations

- It is rare for nonlinear equations to have a **closed form solution** that can be found through algebraic manipulation

General solution for $ax^2 + bx + c = 0$:

$$x = \frac{-b \pm \sqrt{b^2 - 4ac}}{2a}$$

(assuming we know how take square roots of a negative number $b^2 - 4ac < 0 \dots$)

Closed form solutions of nonlinear equations

- General solution for $ax^3 + bx^2 + cx + d = 0$:

$$\begin{aligned}
 x_1 &= -\frac{b}{3a} \\
 &\quad -\frac{1}{3a}\sqrt[3]{\frac{1}{2}\left[2b^3 - 9abc + 27a^2d + \sqrt{(2b^3 - 9abc + 27a^2d)^2 - 4(b^2 - 3ac)^3}\right]} \\
 &\quad -\frac{1}{3a}\sqrt[3]{\frac{1}{2}\left[2b^3 - 9abc + 27a^2d - \sqrt{(2b^3 - 9abc + 27a^2d)^2 - 4(b^2 - 3ac)^3}\right]} \\
 x_2 &= -\frac{b}{3a} \\
 &\quad +\frac{1+i\sqrt{3}}{6a}\sqrt[3]{\frac{1}{2}\left[2b^3 - 9abc + 27a^2d + \sqrt{(2b^3 - 9abc + 27a^2d)^2 - 4(b^2 - 3ac)^3}\right]} \\
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 x_3 &= -\frac{b}{3a} \\
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 \end{aligned}$$

Numerical methods

- ▶ It is rare for nonlinear equations to have a **closed form solution** that can be found through algebraic manipulation
- ▶ We require numerical (approximate) solution
- ▶ This may still be tricky for large systems of nonlinear equations

Newton's Method for finding the zeros of $F(x)$

- ▶ Assume we have access to the **differentiable** function $F(x)$ and its derivative $F'(x)$.
- ▶ Assume we have **one initial point x_0** that is close to the **unknown zero x^*** such that $F(x^*) = 0$.
- ▶ We look for a sequence of iterates x_0, x_1, \dots such that

$$\lim_{n \rightarrow \infty} F(x_n) \rightarrow 0.$$

i.e. the sequence converges to x^* . In practice, we only take finitely many **iterations**. This is called an **iterative method**.

Derivation of Newton' Method

- ▶ Since F is differentiable, we can write its tangent at x_0 as:

$$T_0(x) = F(x_0) + F'(x_0)(x - x_0).$$

- ▶ Find the intersection between the tangent and the x -axis as:

$$F(x_0) + F'(x_0)(x_1 - x_0) = 0$$

or after manipulation, $x_1 = x_0 - \frac{F(x_0)}{F'(x_0)}$.

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- ▶ By repeating the tangent approximation at every step we get:

$$x_{n+1} = x_n - \frac{F(x_n)}{F'(x_n)}, \quad n = 1, 2, \dots$$

a sequence of iterates x_n that converges to x^* **under certain conditions** (to be discussed in detail later).

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Pros and cons

Performance

- ▶ Fast (quadratic convergence rate)
- ▶ Not robust

Other issues

- ▶ Requires the derivative function
- ▶ Requires a "good" initial guess