Computer Science Tripos
Syllabus and Booklist 2019–20

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**Introduction to Part IA**

**Entry to the Computer Science Tripos**

The only essential GCE A level for admission to Cambridge to read for the Computer Science Tripos is Mathematics. Also desirable are Further Mathematics and a physical science (Physics, Chemistry or Geology) at A level, or at AS level if not taken at A level. Some colleges may ask candidates to take the Advanced Extension Award or STEP papers in Mathematics.

**Computer Science Tripos Part IA**

Part IA students taking the 75% *Computer Science option* will attend all lectures for Papers 1, 2 and 3. In addition they attend the Mathematics course offered for Part IA of the Natural Sciences Tripos (NST).

Students taking the 50% *Computer Science option* will take one of the following:

- Part IA students accepted to read *Computer Science with Mathematics* will attend the lectures for Papers 1 and 2 of the Computer Science Tripos in addition to Papers 1 and 2 of Part IA of the Mathematical Tripos.
- Part IA students who take a Natural Science option selected from Chemistry, Evolution and Behaviour, Earth Sciences, Physics, and Physiology of Organisms will attend Papers 1 and 2 of the Computer Science Tripos as well as the Mathematics course offered for Part IA of the Natural Sciences Tripos (NST).

An A level in a science subject is desirable for students taking an NST option. Computer Science students taking an NST option are expected to undertake practical work on the same basis as for the Natural Science Tripos.

**Natural Sciences Part IA students**

There is a Computer Science option in the first year of the Natural Sciences Tripos, counting as one quarter of the year’s work. Students taking this option attend all the lectures and practicals for Paper 1.

**The curriculum**

This document lists the courses offered by the Computer Laboratory for Papers 1, 2 and 3 of Part IA of the Computer Science Tripos. Separate booklets give details of the syllabus for the second- and third-year courses in Computer Science.

The syllabus information given here is for guidance only and should not be considered definitive. Current timetables can be found at [http://www.cl.cam.ac.uk/teaching/timetables/](http://www.cl.cam.ac.uk/teaching/timetables/)
For most of the courses listed below, a list of recommended books is given. These are roughly in order of usefulness, and lecturers have indicated by means of an asterisk those books which are most recommended for purchase by College libraries.

The Computer Laboratory Library aims to keep at least one copy of each of the course texts in “The Booklocker” (see [http://www.cl.cam.ac.uk/library/](http://www.cl.cam.ac.uk/library/)).

For further copies of this booklet and for answers to general enquiries about Computer Science courses, please get in touch with:

Teaching Administrator  
University of Cambridge  
Computer Laboratory  
William Gates Building  
J J Thomson Avenue  
Cambridge  
CB3 0FD

telephone: 01223 763505  
fax: 01223 334678  
e-mail: teaching-admin@cl.cam.ac.uk
Michaelmas Term 2019: Part IA lectures

Paper 1: Foundations of Computer Science

Lecturers: Professor A. Mycroft, Dr A. Prorok

No. of lectures and practicals: 12 + 5 (NST students will take 4 practicals)

Suggested hours of supervisions: 3

This course is a prerequisite for Programming in Java and Prolog (Part IB).

Aims

The main aim of this course is to present the basic principles of programming. As the introductory course of the Computer Science Tripos, it caters for students from all backgrounds. To those who have had no programming experience, it will be comprehensible; to those experienced in languages such as C, it will attempt to correct any bad habits that they have learnt.

A further aim is to introduce the principles of data structures and algorithms. The course will emphasise the algorithmic side of programming, focusing on problem-solving rather than on hardware-level bits and bytes. Accordingly it will present basic algorithms for sorting, searching, etc., and discuss their efficiency using $O$-notation. Worked examples (such as polynomial arithmetic) will demonstrate how algorithmic ideas can be used to build efficient applications.

The course will use a functional language (ML). ML is particularly appropriate for inexperienced programmers, since a faulty program cannot crash and ML’s unobtrusive type system captures many program faults before execution. The course will present the elements of functional programming, such as curried and higher-order functions. But it will also introduce traditional (procedural) programming, such as assignments, arrays and references.

Lectures

- **Introduction to Programming.** The role of abstraction and representation. Introduction to integer and floating-point arithmetic. Declaring functions. Decisions and booleans. Example: integer exponentiation.


- **Lists.** Basic list operations. Append. Naïve versus efficient functions for length and reverse. Strings.

- **More on lists.** The utilities `take` and `drop`. Pattern-matching: zip, unzip. A word on polymorphism. The “making change” example.
• **Sorting.** A random number generator. Insertion sort, mergesort, quicksort. Their efficiency.

• **Datatypes and trees.** Pattern-matching and case expressions. Exceptions. Binary tree traversal (conversion to lists): preorder, inorder, postorder.


• **Functions as values.** Nameless functions. Currying. The “apply to all” functional, map. *Examples:* matrix transpose and product. The predicate functionals filter and exists.

• **Sequences, or lazy lists.** Non-strict functions such as IF. Call-by-need *versus* call-by-name. Lazy lists. Their implementation in ML. Applications, for example Newton-Raphson square roots.


• **Polynomial arithmetic.** Addition, multiplication of polynomials using ideas from sorting, etc.

• **Elements of procedural programming.** Address *versus* contents. Assignment *versus* binding. Own variables. Arrays, mutable or not. Introduction to linked lists.

**Objectives**

At the end of the course, students should

• be able to write simple ML programs;

• understand the fundamentals of using a data structure to represent some mathematical abstraction;

• be able to estimate the efficiency of simple algorithms, using the notions of average-case, worse-case and amortised costs;

• know the comparative advantages of insertion sort, quick sort and merge sort;

• understand binary search and binary search trees;

• know how to use currying and higher-order functions;

• understand how ML combines imperative and functional programming in a single language.
Recommended reading


For reference only:

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Paper 1: Object-Oriented Programming

*Lecturer: Dr A.C. Rice*

*No. of lectures and practicals: 12 + 5*

*Suggested hours of supervisions: 3*

**Aims**

The goal of this course is to provide students with an understanding of Object-Oriented Programming. Concepts are demonstrated in multiple languages, but the primary language is Java.

**Lecture syllabus**

- **Types, Objects and Classes** Moving from functional to imperative. Functions, methods. Control flow. values, variables and types. Primitive Types. Classes as custom types. Objects vs Classes. Class definition, constructors. Static data and methods.


- **Polymorphism and Multiple Inheritance** Polymorphism in ML and Java. Multiple inheritance. Interfaces in Java.


**Copying Objects.** Shallow and deep copies. Copy constructors. Cloning in Java. Cloneable as a marker interface in Java.

**Design Language evolution** Need for languages to evolve. Generics in Java. Type erasure. Introduction to Java 8: Lambda functions, functions as values, method references, streams.

**Design Patterns** Introduction to design patterns. Open-closed principle. Examples of Singleton, Decorator, State, Composite, Strategy, Observer. [2 lectures]

**Objectives**

At the end of the course students should

- be familiar with the main features and limitations of the Java language;
- be able to write a Java program to solve a well specified problem;
- understand the principles of OOP;
- be able to demonstrate good object-oriented programming skills in Java;
- be able to describe, recognise, apply and implement selected design patterns in Java;
- be familiar with common errors in Java and its associated libraries;
- understand a Java program written by someone else;
- be able to debug and test Java programs;
- be familiar with major parts of Java 8 SE libraries;
- understand how to read Javadoc library documentation and reuse library code.

**Recommended reading**

No single text book covers all of the topics in this course. For those new to OOP, the best introductions are usually found in the introductory programming texts for OOP languages (such as Java, python or C++). Look for those that are for people new to programming rather than those that are designed for programmers transitioning between languages (the Deitel book is highlighted for this reason). The web is also a very useful resource — look for Java tutorials.
Paper 2: Digital Electronics

This course is not taken by NST students.

Lecturer: Dr I.J. Wassell

No. of lectures and practical classes: 12 + 7

Suggested hours of supervisions: 3

This course is a prerequisite for Operating Systems and Computer Design (Part IB), ECAD and Architecture Practical Classes (Part IB).

Aims

The aims of this course are to present the principles of combinational and sequential digital logic design and optimisation at a gate level. The use of n and p channel MOSFETs for building logic gates is also introduced.

Lectures


- **Logic minimisation.** Truth tables and normal forms. Karnaugh maps. Quine-McCluskey method.

- **Binary adders.** Half adder, full adder, ripple carry adder, fast carry generation.

- **Combinational logic design: further considerations.** Multilevel logic. Gate propagation delay. An introduction to timing diagrams. Hazards and hazard elimination. Other ways to implement combinational logic.

- **Introduction to practical classes.** Prototyping box. Breadboard and Dual in line (DIL) packages. Wiring. Use of oscilloscope.


- **Sequential logic.** Counters: Ripple and synchronous. Shift registers.
• **Synchronous State Machines.** Moore and Mealy finite state machines (FSMs). Reset and self starting. State transition diagrams. Elimination of redundant states.

• **Further state machines.** State assignment: sequential, sliding, shift register, one hot. Implementation of FSMs.

• **Electronics, Devices and Circuits.** Current and voltage, resistance, basic circuit theory, the potential divider. Solving non-linear circuits. Materials, semiconductors and the p-n junction, i.e., the diode. n and p channel MOSFETs and n-MOSFET logic, e.g., n-MOSFET inverter. Switching speed and power consumption problems in n-MOSFET logic. CMOS logic. Logic families. Noise margin. Analogue interfacing and operational amplifiers. [3 lectures]

**Objectives**

At the end of the course students should

- understand the relationships between combination logic and boolean algebra, and between sequential logic and finite state machines;
- be able to design and minimise combinational logic;
- appreciate tradeoffs in complexity and speed of combinational designs;
- understand how state can be stored in a digital logic circuit;
- know how to design a simple finite state machine from a specification and be able to implement this in gates and edge triggered flip-flops;
- understand how to use MOSFETs to build digital logic circuits.
- understand the effect of finite load capacitance on the performance of digital logic circuits.
- understand basic analogue interfacing.

**Recommended reading**


Books for reference:
Paper 2: Discrete Mathematics

This course is not taken by NST students.

Lecturers: Professor G. Winskel and Professor F. Stajano

No. of lectures: 24 (continued into Lent term)

Suggested hours of supervisions: 6

This course is a prerequisite for all theory courses as well as: Probability, Security, Artificial Intelligence, Compiler Construction and the following Part II courses: Machine Learning and Bayesian Inference and Cryptography

Aims

The course aims to introduce the mathematics of discrete structures, showing it as an essential tool for computer science that can be clever and beautiful.

Lectures

- **Proof [5 lectures].**
  

- **Numbers [5 lectures].**
  
  Number systems: natural numbers, integers, rationals, modular integers. The Division Theorem and Algorithm. Modular arithmetic. Sets: membership and comprehension. The greatest common divisor, and Euclid’s Algorithm and Theorem. The Extended Euclid’s Algorithm and multiplicative inverses in modular arithmetic. The Diffie-Hellman cryptographic method. Mathematical induction: Binomial Theorem, Pascal’s Triangle, Fundamental Theorem of Arithmetic, Euclid’s infinity of primes.

- **Sets [9 lectures].**
  

- **Formal languages and automata [5 lectures].**
  Introduction to inductive definitions using rules and proof by rule induction. Abstract syntax trees.
  Regular expressions and their algebra.
  Finite automata and regular languages: Kleene’s theorem and the Pumping Lemma.

**Objectives**

On completing the course, students should be able to

- prove and disprove mathematical statements using a variety of techniques;
- apply the mathematical principle of induction;
- know the basics of modular arithmetic and appreciate its role in cryptography;
- understand and use the language of set theory in applications to computer science;
- define sets inductively using rules and prove properties about them;
- convert between regular expressions and finite automata;
- use the Pumping Lemma to prove that a language is not regular.

**Recommended reading**

Paper 3: Databases

This course is only taken by Part IA and Part IB Paper 3 students

Lecturer: Dr T.G. Griffin

No. of lectures and practical classes: 8 + 4

Suggested hours of supervisions: 3

Prerequisite courses: None

Aims

This course introduces basic concepts for database systems as seen from the perspective of application designers. That is, the focus is on the abstractions supported by database management systems and not on how those abstractions are implemented.

The database world is currently undergoing swift and dramatic transformations largely driven by Internet-oriented applications and services. Today many more options are available to database application developers than in the past and so it is becoming increasingly difficult to sort fact from fiction. The course attempts to cut through the fog with a practical approach that emphasises engineering tradeoffs that underpin these recent developments and also guide our selection of “the right tool for the job.”

This course covers three approaches. First, the traditional mainstay of the database industry — the relational approach — is described with emphasis on eliminating logical redundancy in data. Then two representatives of recent trends are presented — graph-oriented and document-oriented databases. The lectures are tightly integrated with the associated practical sessions where students gain hands-on experience with all three of these approaches.

Lectures

- **Introduction.** What is a database system? What is a data model? A central tradeoff in the choice of data representation: optimise for ease of updating or for fast query response. On-Line Transaction Processing (OLTP) versus On-line Analytical Processing (OLAP). Application independent versus application specific data representations. [1 lecture]

- **Conceptual modeling** The Entity-Relationship (ER) approach as an implementation-independent technique for modeling data. [1 lecture]

- **The relational model** Implementing ER models with relational tables. Relational algebra and SQL. Update anomalies caused by logical redundancy. Minimise logical redundancy with normalised data representation. What is transitive closure? Why SQL struggles with transitive closure. [2 lectures]

- **The graph-oriented model** The NoSQL movement. Implementing ER models in a graph-oriented database. Graph databases: optimised for computing transitive closure. Path-oriented queries. [2 lectures]

• **The multi-dimensional model**. Data cubes, star schema, data warehouse. [1 lecture]

**Objectives**

At the end of the course students should

• be able to design entity-relationship diagrams to represent simple database application scenarios

• know how to convert entity-relationship diagrams to relational- and graph-oriented implementations

• understand the fundamental tradeoff between the ease of updating data and the response time of complex queries

• understand that no single data architecture can be used to meet all data management requirements

• be familiar with recent trends in the database area.

**Recommended reading**


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**Paper 3: Introduction to Graphics**

This course is only taken by Part IA and Part IB Paper 3 students.

*Lecturer: Dr R.K. Mantiuk and Professor P. Robinson*

*No. of lectures and practical classes: 8 + 4*

*Suggested hours of supervisions: 2*

*Prerequisite courses: None*

*This course is a prerequisite for Further Graphics*

**Aims**

To introduce the necessary background, the basic algorithms, and the applications of computer graphics and graphics hardware.
Lectures

- **Background.** What is an image? Resolution and quantisation. Storage of images in memory. [1 lecture]


- **Colour.** Perception of colour. Colour spaces. [1 lecture]

Objectives

By the end of the course students should be able to:

- understand and apply in practice basic concepts of ray-tracing: ray-object intersection, reflections, refraction, shadow rays, distributed ray-tracing, direct and indirect illumination;

- describe and explain the following algorithms: Gouraud and Phong shading, z-buffer, texture mapping, double buffering, mip-map, bump- and normal-mapping;

- use matrices and homogeneous coordinates to represent and perform 2D and 3D transformations; understand and use 3D to 2D projection, the viewing volume, and 3D clipping;

- implement OpenGL code for rendering of polygonal objects, control camera and lighting, work with vertex and fragment shaders;

- describe a number of colour spaces and their relative merits.

Recommended reading

Scientific Computing Practical Course

This is a practical course taken by Part IA CST students only. It is taught through one introductory lecture, followed by an online course which is equivalent in content to approximately 5 lectures. Students will normally work through the online component at their own pace. The course will be entirely examined through practical exercises.

Lecturer: Dr D. Wischik

No. of lectures: 1 (plus an online course with roughly 5 lectures worth of material in the Lent term)

Suggested hours of supervisions: none

Prerequisite courses: Foundations of Computer Science, NST Mathematics

This course is a prerequisite for Foundations of Data Science (Part IB)

Aims

This course is a hands-on introduction to using computers to investigate scientific models and data.

Syllabus

- Python notebooks. Overview of the Python programming language. Use of notebooks for scientific computing.
- Working with data. Data import. Common ways to summarize and plot data, for univariate and multivariate analysis.

Objectives

At the end of the course students should

- be able to import data, plot it, and summarize it appropriately
- be able to write fast vectorized code for scientific / data work
Lent Term 2020: Part IA lectures

Paper 1: Algorithms

Lecturers: Professor F. Stajano and Dr D. Wischik

No. of lectures and practical classes: 24 + 3 (NST students take 1 practical)

Suggested hours of supervisions: 6

Prerequisite courses: Foundations of Computer Science, Object-Oriented Programming

This course is a prerequisite for: Artificial Intelligence, Complexity Theory, Further Graphics, Prolog and the following Part II courses: Advanced Algorithms and Machine Learning and Bayesian Inference

Aims

The aim of this course is to provide an introduction to computer algorithms and data structures, with an emphasis on foundational material.

Lectures

• **Sorting.** Review of complexity and O-notation. Trivial sorting algorithms of quadratic complexity. Review of merge sort and quicksort, understanding their memory behaviour on statically allocated arrays. Heapsort. Stability. Other sorting methods including sorting in linear time. Median and order statistics. [Ref: CLRS3 chapters 1, 2, 3, 6, 7, 8, 9] [about 4 lectures]

• **Strategies for algorithm design.** Dynamic programming, divide and conquer, greedy algorithms and other useful paradigms. [Ref: CLRS3 chapters 4, 15, 16] [about 3 lectures]

• **Data structures.** Elementary data structures: pointers, objects, stacks, queues, lists, trees. Binary search trees. Red-black trees. B-trees. Hash tables. Priority queues and heaps. [Ref: CLRS3 chapters 6, 10, 11, 12, 13, 18] [about 5 lectures]


• **Advanced data structures.** Binomial heap. Amortized analysis: aggregate analysis, potential method. Fibonacci heaps. Disjoint sets. [Ref: CLRS3 chapters 17, 19, 20, 21] [about 4 lectures]

• **Geometric algorithms.** Intersection of segments. Convex hull: Graham’s scan, Jarvis’s march. [Ref: CLRS3 chapter 33] [about 1 lecture]
Objectives

At the end of the course students should:

- have a thorough understanding of several classical algorithms and data structures;
- be able to analyse the space and time efficiency of most algorithms;
- have a good understanding of how a smart choice of data structures may be used to increase the efficiency of particular algorithms;
- be able to design new algorithms or modify existing ones for new applications and reason about the efficiency of the result.

Recommended reading


Paper 2: Operating Systems

This course is not taken by NST students.

Lecturer: Dr. R. Mortier

No. of lectures: 12

Suggested hours of supervisions: 3

Prerequisite courses: Pre-arrival course, Digital Electronics

This course is a prerequisite for Concurrent & Distributed Systems (Part IB), Security and Mobile and Sensor Systems (Part II).

Aims

The overall aim of this course is to provide a general understanding of the structure and key functions of the operating system. Case studies will be used to illustrate and reinforce fundamental concepts.

Lectures

- **Introduction to operating systems.** Abstract view of an operating system. Elementary computer architecture. OS evolution: multi-programming, time-sharing. [1 lecture]

• **Processes.** Job/process concepts. Lifecycle. Process management. Inter-process communication. [1 lecture]

• **Scheduling.** Scheduling basics: CPU-I/O interleaving, (non-)preemption, context switching. Scheduling algorithms: FCFS, SJF, SRTF, priority scheduling, round robin. Combined schemes. [2 lectures]


• **I/O subsystem.** General structure. Polled mode versus interrupt-driven I/O. Application I/O interface: block and character devices, buffering, blocking versus non-blocking I/O. Other issues: caching, scheduling, spooling, performance. [1 lecture]


**Objectives**

At the end of the course students should be able to

• describe the general structure and purpose of an operating system;

• explain the concepts of process, address space, and file;

• compare and contrast various CPU scheduling algorithms;

• understand the differences between segmented and paged memories, and be able to describe the advantages and disadvantages of each;

• compare and contrast polled, interrupt-driven and DMA-based access to I/O devices.
Recommended reading

* Bacon, J. & Harris, T. (2003). *Operating systems*. Addison-Wesley (3rd ed.).
Leffler, S. (1989). *The design and implementation of the 4.3BSD Unix operating system*. Addison-Wesley.

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Paper 3: Machine Learning and Real-world Data

This course is only taken by Part IA and Part IB Paper 3 students.

Lecturers: Professor S.H. Teufel, Dr P. Buttery and Professor A. Copestake

No. of lectures and practical classes: 16

Suggested hours of supervisions: 4

Prerequisite courses: NST Mathematics

Aims

This course introduces students to machine learning algorithms as used in real-world applications, and to the experimental methodology necessary to perform statistical analysis of large-scale data from unpredictable processes. Students will perform 3 extended practicals, as follows:

- Statistical classification: Determining movie review sentiment using Naive Bayes (7 sessions);
- Sequence Analysis: Hidden Markov Modelling and its application to a task from biology (predicting protein interactions with a cell membrane) (4 sessions);
- Analysis of social networks, including detection of cliques and central nodes (5 sessions).

Syllabus

- **Topic One: Statistical Classification [7 sessions]**.
  Introduction to sentiment classification.
  Naive Bayes parameter estimation.
  Statistical laws of language.
  Statistical tests for classification tasks.
  Cross-validation and test sets.
  Uncertainty and human agreement.
• **Topic Two: Sequence Analysis [4 sessions].**
  Hidden Markov Models (HMM) and HMM training.
  The Viterbi algorithm.
  Using an HMM in a biological application.

• **Topic Three: Social Networks [5 sessions].**
  Properties of networks: Degree, Diameter.
  Betweenness Centrality.
  Clustering using betweenness centrality.

**Objectives**

By the end of the course students should be able to:

• understand and program two simple supervised machine learning algorithms;

• use these algorithms in statistically valid experiments, including the design of baselines, evaluation metrics, statistical testing of results, and provision against overtraining;

• visualise the connectivity and centrality in large networks;

• use clustering (i.e., a type of unsupervised machine learning) for detection of cliques in unstructured networks.

**Recommended reading**

Jurafsky, D. & Martin, J. (2008). *Speech and language processing*. Prentice Hall.
Easter Term 2020: Part IA lectures

Introduction to Probability

Lecturer: Prof. M. Jamnik, Dr T. Sauerwald
No. of lectures: 8
This course is a prerequisite for Probability and Computation (Part II).

Aims

The aim of the course is to introduce the theory of computational complexity. The course will explain measures of the complexity of problems and of algorithms, based on time and space used on abstract models. Important complexity classes will be defined, and the notion of completeness established through a thorough study of NP-completeness. Applications to cryptographic protocols will be considered.

Lectures

Part 1 - Introduction to Probability

- **Introduction.** Counting/Combinatorics (revision), Probability Space, Axioms, Union Bound.

- **Conditional probability.** Conditional Probabilities and Independence, Bayes’ Theorem, Partition Theorem

Part 2 - Discrete Random Variables


- **Probability distributions.** Definition and Properties of Expectation, Variance, different ways of computing them, Examples of important Distributions (Bernoulli, Binomial, Geometric, Poisson), Primer on Continuous Distributions including Normal and Exponential Distributions.

- **Multivariate distributions.** Multiple Random Variables, Joint and Marginal Distributions, Independence of Random Variables, Covariance.

Part 3 - Moments and Limit Theorems

- **Introduction.** Law of Average, Useful inequalities (Markov and Chebyshev), Weak Law of Large Numbers (including Proof using Chebyshev’s inequality), Examples.
• **Moments and Central Limit Theorem.** Introduction to Moments of Random Variables, Central Limit Theorem (Proof using Moment Generating functions), Example.

Part 4 - Applications/Statistics


• **Algorithms.** Online Algorithms (Secretary Problem, Odd's Algorithm).

**Objectives**

At the end of the course students should

• understand the basic principles of probability spaces and random variables
• be able to formulate problems using concepts from probability theory and compute or estimate probabilities
• be familiar with more advanced concepts such as moments, limit theorems and applications such as parameter estimation

**Recommended reading**


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**Paper 2: Software and Security Engineering**

*Lecturer: Dr. A.R. Beresford*

*No. of lectures: 11*

*Suggested hours of supervisions: 3*

*This course is a prerequisite for the Group Project.*

**Aims**

This course aims to introduce students to software and security engineering, and in particular to the problems of building large systems, safety-critical systems and systems that must withstand attack by capable opponents. Case histories of failure are used to illustrate what can go wrong, and current software and security engineering practice is studied as a guide to how failures can be avoided.
Lectures

- **What is a security policy or a safety case?** Definitions and examples; one-way flows for both confidentiality and safety properties; separation of duties. Top-down and bottom-up analysis methods. What architecture can do, versus benefits of decoupling policy from mechanism.

- **Predicting user behaviour.** Predicting and mitigating user errors. The hierarchy of harms. Attitudes to risk: expected utility, prospect theory, framing, status quo bias, gender. The characteristics of human memory; forgetting passwords versus guessing them.

- **Security protocols;** how to enforce policy using cryptography and structured human interaction. Man-in-the-middle attacks. The role of verification and its limitations.

- **Bugs of different types:** design errors, implementation errors affecting arithmetic, logic, syntax, and concurrency. Code injection attacks. Defensive programming: secure coding, contracts. Fuzzing.

- **The software crisis.** Examples of large-scale project failure, such as the London Ambulance Service system and the NHS National Programme for IT. Intrinsic difficulties with complex software.

- **The software life cycle.** The software life cycle. Getting the specification right; requirements analysis methods; modular design; the role of prototyping; the waterfall and spiral models.

- **Designing for Testability.** Identifying different types of tests and how to use them effectively. Writing a good unit test; understanding techniques for measuring test quality.

- **Modern development practice and quality assurance.** Tools to support code management, code review and test case generation. Continuous integration, refactoring, release engineering, patch strategies. The need for code indexing, code ownership, library management, design documentation and the maintenance of safety and security ratings.

- **Software-as-a-Service.** The Software-as-a-Service architecture, including a/b testing, phased release of client and server components and rearchitecting systems while in operation.

- **Critical systems:** where real-time performance, safety or security is essential. Examples of catastrophic failure; problems with usability and human error for both safety engineering and security engineering.

- **Real-world challenges in combining safety and security.** Project planning tools; PERT and GANTT charts. Open source: advantages and drawbacks. Evaluation and assurance; maintaining a security rating or a safety case.
Objectives

At the end of the course students should know how writing programs with tough assurance targets, in large teams, or both, differs from the programming exercises they have engaged in so far. They should understand the different models of software development described in the course as well as the value of various development and management tools. They should understand the development life cycle and its basic economics. They should understand the various types of bugs, vulnerabilities and hazards, how to find them, and how to avoid introducing them. Finally, they should be prepared for the organizational aspects of their Part IB group project.

Recommended reading


Further reading:

[http://www.cs.ucl.ac.uk/staff/A.Finkelstein/las.html](http://www.cs.ucl.ac.uk/staff/A.Finkelstein/las.html)
Paper 3: Interaction Design

This course is only taken by Part IA and Part IB Paper 3 students.

Lecturer: Dr H. Gunes

No. of lectures and practical classes: 8 + 8

Suggested hours of supervisions: 2

Prerequisite courses: Java

This course is a prerequisite for Human-Computer Interaction (Part II)

Aims

The aim of this course is to provide an introduction to interaction design, with an emphasis on understanding and experiencing the user interface design process from requirements and data gathering to implementation and evaluation, while gaining an understanding of the background to human factors. This course focuses equally on design and implementation.

Lectures

- **Course overview and requirements analysis.** Introduction to the course and the practicals. Identifying potential users and understanding their tasks. Identifying and establishing non-functional and functional requirements.

- **Data gathering.** Data collection and quantitative/qualitative analysis techniques.


- **Case studies from the industry.** Guest lecture (the schedule of this lecture is subject to change).

- **Cognitive aspects.** Attention, perception/recognition, memory, context and grouping, and their implications for interaction design. Cognitive frameworks.

- **Evaluation.** Introduction, evaluation techniques, and an evaluation case study.

- **Student (group) presentations**

Objectives

By the end of the course students should

- have a thorough understanding of the iterative design process and be able to apply it to interaction design;
• be able to design new user interfaces that are informed by principles of human visual perception and cognition;

• be able to construct user interfaces using Java with a strong emphasis on users, usability and appearance;

• be able to evaluate existing or new user interfaces using multiple techniques;

• be able to compare and contrast different design techniques and to critique their applicability to new domains.

Recommended reading


Preparing to Study Computer Science

For general advice about preparing for the Computer Science course at Cambridge and for details of the pre-arrival course, please see: http://www.cl.cam.ac.uk/freshers/
Introduction to Part IB

This document lists the courses offered by the Computer Laboratory for Part IB of the Computer Science Tripos. Separate booklets give details of the syllabus for other Parts of the Computer Science Tripos.

Some courses are specific to either Paper 3 or Paper 7, and have been marked as such in this booklet. Those students following the 75% Computer Science option, who have taken Paper 3 in Part IA, will attend lectures for the Paper 7 courses. Those students taking the 50% Computer Science option, will attend the Paper 3 courses instead. The remaining courses are taken by all Part IB students.

The syllabus information given here is for guidance only and should not be considered definitive. Current timetables can be found at http://www.cl.cam.ac.uk/teaching/timetables/

For most of the courses listed below, a list of recommended books is given. These are roughly in order of usefulness, and lecturers have indicated by means of an asterisk those books which are most recommended for purchase by College libraries.

The Computer Laboratory Library aims to keep at least one copy of each of the course texts in “The Booklocker” (see http://www.cl.cam.ac.uk/library/).

For copies of the other syllabus booklets and for answers to general enquiries about Computer Science courses, please get in touch with:

Teaching Administrator
University of Cambridge
Computer Laboratory
William Gates Building
J J Thomson Avenue
Cambridge
CB3 0FD

telephone: 01223 763505
fax: 01223 334678
e-mail: teaching-admin@cl.cam.ac.uk
Michaelmas Term 2019: Part IB lectures

Computer Design

Lecturers: Professor S.W. Moore and Dr T.M. Jones
No. of lectures: 18 (plus 4 via a web-based tutor)
Suggested hours of supervisions: 5
Prerequisite course: Digital Electronics
Companion course: Electronic Computer Aided Design (ECAD)
This course is a prerequisite for the Part II course Comparative Architectures.

Aims

The aims of this course are to introduce a hardware description language (SystemVerilog) and computer architecture concepts in order to design computer systems. The parallel ECAD+Arch practical classes will allow students to apply the concepts taught in lectures. The course starts with a web-based SystemVerilog tutor which is a prerequisite for the ECAD+Arch practical classes. There are then eighteen lectures in three six-lecture parts. Part 1 goes from gates to a simple processor. Part 2 looks at instruction set and computer architecture. Part 3 analyses the architecture of modern systems-on-chip.

Lectures

Part 0 - SystemVerilog Web tutor

- This web tutor is a prerequisite to starting the ECAD+Arch laboratory sessions [equivalent to approximately 4 lectures]

Part 1 - Gates to processors [lecturer: Simon Moore]

- Introduction and motivation. [1 lecture] Current technology, technology trends, ECAD trends, challenges.
- SystemVerilog FPGA design. [1 lecture] Practicalities of mapping SystemVerilog descriptions of hardware (including a processor) onto an FPGA board. Tips and pitfalls when generating larger modular designs.
- Chip, board and system testing. [1 lecture] Production testing, fault models, testability, fault coverage, scan path testing, simulation models.
- Building a simple computer. [2 lectures]
Part 2 - Instruction sets and introduction to computer architecture [lecturer: Simon Moore]

- **Historical perspective on computer architecture.** [1 lecture] EDSAC versus Manchester Mark I.
- **RISC machines.** [1 lecture] Introduction to ARM and MIPS RISC processor designs.
- **CISC and virtual machines** [1 lecture] The Intel x86 instruction set and the Java Virtual Machine (JVM).
- **Memory hierarchy.** [1 lecture] Caching, etc.
- **Hardware support for operating systems.** [1 lecture] Memory protection, exceptions, interrupts, etc.
- **Pipelining and data paths.** [1 lecture]

Part 3 - Systems-on-chip [lecturer: Timothy Jones]

- **Overview of Systems-on-Chip (SoCs) and DRAM.** [1 lecture] High-level SoCs, DRAM storage and accessing.
- **Multicore Processors.** [2 lectures] Communication, cache coherence, barriers and synchronisation primitives.
- **Graphics processing units (GPUs)** [2 lectures] Basic GPU architecture and programming.
- **Future Directions** [1 lecture] Where is computer architecture heading?

**Objectives**

At the end of the course students should

- be able to read assembler given a guide to the instruction set and be able to write short pieces of assembler if given an instruction set or asked to invent an instruction set;
- understand the differences between RISC and CISC assembler;
- understand what facilities a processor provides to support operating systems, from memory management to software interrupts;
- understand memory hierarchy including different cache structures and coherency needed for multicore systems;
- understand how to implement a processor in SystemVerilog;
- appreciate the use of pipelining in processor design;
- have an appreciation of control structures used in processor design;
• have an appreciation of system-on-chips and their components;
• understand how DRAM stores data;
• understand how multicore processors communicate;
• understand how GPUs work and have an appreciation of how to program them.

Recommended reading


Recommended further reading:


Pointers to sources of more specialist information are included in the lecture notes and on the associated course web page.

Concurrent and Distributed Systems

*Lecturer: Dr R.M. Mortier, Dr E. Kalyvianaki and Dr A. Madhavapeddy*

No. of lectures: 16

Suggested hours of supervisions: 4

Prerequisite courses: Operating Systems, Object-Oriented Programming

This course is a pre-requisite for Mobile and Sensor Systems (Part II).

Aims

This course considers two closely related topics, Concurrent Systems and Distributed Systems, over 16 lectures. The aim of the first half of the course is to introduce concurrency control concepts and their implications for system design and implementation. The aims of the latter half of the course are to study the fundamental characteristics of distributed systems, including their models and architectures; the implications for software design; some of the techniques that have been used to build them; and the resulting details of good distributed algorithms and applications.

Lectures: Concurrency

• **Introduction to concurrency, threads, and mutual exclusion** Introduction to concurrent systems; threads; interleaving; preemption; parallelism; execution
orderings; processes and threads; kernel vs. user threads; M:N threads; atomicity; mutual exclusion; and mutual exclusion locks (mutexes).

- **Further mutual exclusion, semaphores, producer-consumer, and MRSW**
  Hardware foundations for atomicity; locks and invariants; semaphores; condition synchronisation; N-resource allocation; two-party and generalised producer-consumer; Multi-Reader, Single-Write (MRSW) locks.

- **CCR, monitors, and concurrency in practice**
  Conditional critical regions (CCR); monitors; condition variables; signal-wait vs. signal-continue semantics; concurrency in practice (kernels, pthreads, Java).

- **Safety and liveness**
  Safety vs. liveness; deadlock; the Dining Philosophers; resource allocation graphs; deadlock prevention, avoidance, detection, and recovery; livelock; priority inversion; priority inheritance.

- **Concurrency without shared data; transactions**
  Active objects; message passing; tuple spaces; CSP; and actor models. Composite operations; transactions; ACID; isolation; and serialisability.

- **Further transactions**
  History graphs; good and bad schedules; isolation vs. strict isolation; 2-phase locking; rollback; timestamp ordering (TSO); and optimistic concurrency control (OCC).

- **Crash recovery, lock-free programming, and transactional memory**
  Write-ahead logging, checkpoints, and recovery. Lock-free programming and software-transactional memory (STM).

- **Concurrent systems case study.**
  Concurrency in the FreeBSD kernel; kernel synchronisation prior to parallelism; Giant-locked kernels; fine-grained locking; primitives and strategies; lock order checking; network-stack work flows; performance scalability; the impact of evolving hardware.

**Lectures: Distributed Systems**

- **Introduction to distributed systems; RPC**
  Advantages and challenges of distributed systems; “middleware”; transparency goals; client-server systems; failures and retry semantics (all-or-nothing; at-most-once; at-least-once). Remote procedure call (RPC); marshalling; interface definition languages (IDLs); SunRPC; external data representation (XDR).

- **Network File System and Object-Oriented Middleware**
  Network File System (NFS); NFSv2; NFSv3; scoping; the implications of a stateless design; performance optimisations. Object-oriented middleware (OOM); Corba ORBs, IDL; DCOM.

- **Practical RPC systems; clocks**
  Remote method invocation (RMI); remote classes vs. serialisable classes; distributed garbage collection; XML-RPC; SOAP and web services; REST. Physical clocks; UTC; computer clocks; clock synchronisation.

- **Clock synchronisation; logical clocks**
  Clock drift and compensation; Cristian’s Algorithm; Berkeley Algorithm; Network Time Protocol (NTP). Logical time, “happens-before”; Lamport clocks; vector clocks.
• **Consistent cuts, process groups, and mutual exclusion** Consistent global state; consistent cuts. Process groups; FIFO ordering; receiving vs. delivering; causal ordering; total ordering. Distributed mutual exclusion; central lock servers; token passing; totally ordered multicast.

• **Elections, consensus, and distributed transactions** Leader elections; ring-based algorithm; the Bully algorithm. Consensus. Distributed transactions; atomic commit protocols; 2-phase commit. Replication and consistency.

• **Replication in distributed systems, CAP, case studies** Replication and consistency (cont); strong consistency; quorum systems; weak consistency; FIFO consistency; eventual consistency; Amazon’s Dynamo; session guarantees; Consistency, Availability and Partitions (CAP); Google datacentre technologies (MapReduce).

• **Further case studies, PubSub, security, NASD/AFS/Coda** Google datacentre technologies (BigTable, Spanner). Access control and the access-control matrix; ACLs vs capabilities; cryptographic capabilities; role-based access control (RBAC); single-system sign-on. NASD, AFS, and Coda.

**Objectives**

At the end of Concurrent Systems portion of the course, students should:

- understand the need for concurrency control in operating systems and applications, both mutual exclusion and condition synchronisation;
- understand how multi-threading can be supported and the implications of different approaches;
- be familiar with the support offered by various programming languages for concurrency control and be able to judge the scope, performance implications and possible applications of the various approaches;
- be aware that dynamic resource allocation can lead to deadlock;
- understand the concept of transaction; the properties of transactions, how they can be implemented, and how their performance can be optimised based on optimistic assumptions;
- understand how the persistence properties of transactions are addressed through logging; and
- have a high-level understanding of the evolution of software use of concurrency in the operating-system kernel case study.

At the end of the Distributed Systems portion of the course, students should:

- understand the difference between simple concurrent systems and distributed systems;
• understand the fundamental properties of distributed systems and their implications for system design;

• understand notions of time synchronisation, including logical clocks, vector clocks, and physical time;

• be familiar with various approaches to data and service replication, as well as the concept of data consistency;

• understand the effects of large scale on the provision of fundamental services and the tradeoffs arising from scale;

• appreciate the implications of individual node and network communications failures on distributed computation;

• be aware of a variety of tools used by distributed-system creators, such as RPC and object-oriented middleware (OOM);

• be familiar with a range of distributed algorithms;

• be familiar with a number of case studies in distributed-system design including: the Network File System (NFS), the Network Time Protocol (NTP), Java Remote Method Invocation (RMI), CORBA, the AFS and Coda filesystems, Network-Attached Secure Disks (NASD), and Google's MapReduce, BigTable, and Spanner systems.

Recommended reading

ECAD and Architecture Practical Classes

Lecturer: Dr. T. Markettos, Professor S.W. Moore & Dr R. Mullins

No. of practical classes: 8

Prerequisite course: Digital Electronics

Companion course: Computer Design

This course is a prerequisite for the Part II course Comparative Architectures.

Aims

The aims of this course are to enable students to apply the concepts learned in the Computer Design course. In particular a web based tutor is used to introduce the SystemVerilog hardware description language, while the remaining practical classes will then allow students to implement the design of components in this language.

Practical Classes

- **Web tutor** This first class uses a web based tutor to rapidly teach the SystemVerilog language (this is part of the lectured component of Computer Design).

- **FPGA design flow** Test driven hardware development for FPGA including an embedded processor and peripherals [3 classes]

- **Embedded system implementation** Embedded system implementation on FPGA [3-4 classes]

Objectives

- Gain experience in electronic computer aided design (ECAD) through learning a design-flow for field programmable gate arrays (FPGAs).

- Learn how to interface to peripherals like a touch screen.

- Learn how to debug hardware and software systems in simulation.

- Understand how to construct and program a heterogeneous embedded system.

Recommended reading


Pointers to sources of more specialist information are included on the associated course web page.
Paper 7: Economics, Law and Ethics

This course is only taken by Part IB and Part II Paper 7 students.

Lecturers: Professor R.J. Anderson and Dr R.N.C. Clayton

No. of lectures: 8

Suggested hours of supervisions: 2

This course is a prerequisite for the Part II courses Business Studies and E-Commerce.

Aims

This course aims to give students an introduction to some basic concepts in economics, law and ethics.

Lectures

- **Game theory.** The choice between cooperation and conflict. Prisoners' Dilemma; Nash equilibrium; hawk–dove; iterated games; evolution of strategies; application to biology and computer science.

- **Classical economics.** Definitions: preference, utility, choice and budget. Pareto efficiency; the discriminating monopolist; supply and demand; elasticity; utility; the marginalist revolution; competitive equilibrium and the welfare theorems. Trade; monopoly rents; public goods; oligopoly.

- **Market failure.** Asymmetric information: the market for lemons; adverse selection; moral hazard; signalling; and brands. Transaction costs and the theory of the firm. Real and virtual networks, supply-side versus demand-side scale economies, Metcalfe's law, the dominant firm model, price discrimination. Behavioural economics: bounded rationality, heuristics and biases.

- **Auctions.** English auctions; Dutch auctions; all-pay auctions; Vickrey auctions. The winner's curse. The revenue equivalence theorem. Mechanism design and the combinatorial auction. Problems with real auctions. Applicability of auction mechanisms in computer science.

- **Principles of law.** Contract and tort; copyright and patent; binding actions; liabilities and remedies; competition law; choice of law and jurisdiction.

- **Law and the Internet.** EU directives including distance selling, electronic commerce, data protection, electronic signatures and copyright; their UK implementation. UK laws that specifically affect the Internet.

- **Ethics.** Philosophies of ethics: authority, intuitionist, egoist and deontological theories. Utilitarian and Rawlsian models. Insights from evolutionary psychology and neorology. The Internet and social policy; current debates on privacy, surveillance, censorship and export control.
Objectives

At the end of the course students should have a basic appreciation of economic and legal terminology and arguments. They should understand some of the applications of economic models to systems engineering and their interest to theoretical computer science. They should also understand the main constraints that markets, legislation and ethics place on firms dealing in information goods and services.

Recommended reading


Further reading:


Foundations of Data Science

*Lecturer: Dr D.J. Wischik*

*No. of lectures and practical classes: 12 + 4*

*Suggested hours of supervisions: 3*

*Prerequisite courses:* either Mathematics for Natural Sciences, or the equivalent from the Maths Tripos

*This course is a prerequisite for:* Part II Machine Learning and Bayesian Inference, Information Retrieval, Quantum Computing, Natural Language Processing.

Aims

This course introduces fundamental tools for describing and reasoning about data. There are two themes: describing the behaviour of random systems; and making inferences based on data generated by such systems. The course will survey a wide range of models and tools, and it will emphasize how to design a model and what sorts of questions one might ask about it.
Lectures


- **Inference.** Estimation, confidence intervals, hypothesis testing, prediction. Bootstrap. Bayesianism. Logistic regression, natural parameters.


Objectives

At the end of the course students should

- be able to formulate basic probabilistic models, including discrete time Markov chains and linear models

- be familiar with common random variables and their uses, and with the use of empirical distributions rather than formulae

- be able to use expectation and conditional expectation, limit theorems, equilibrium distributions

- understand different types of inference about noisy data, including model fitting, hypothesis testing, and making predictions

- understand the fundamental properties of inner product spaces and orthonormal systems, and their application to model representation

Recommended reading


Paper 7: Further Graphics

This course is only taken by Part IB and Part II Paper 7 students.

Lecturer: Dr A. Benton

No. of lectures: 8

Suggested hours of supervisions: 2

Prerequisite courses: Introduction to Graphics

This course is a prerequisite for: Advanced Graphics

Aims

This course explores the modern state of computer graphics, applying long-standing techniques to cutting-edge hardware.

Lectures

The order and content of lectures is provisional and subject to change.

- **Ray Marching.** Signed distance fields and GPU-based realtime rendering. [1 lecture]
- **Computational Geometry.** Mathematics of surfaces. [1 lecture]
- **Bezier curves and NURBS.** These points of data make a beautiful line. [2 lectures]
- **Subdivision Surfaces.** Smooth modeling of continuous surfaces. [1 lecture]
- **OpenGL.** Polygonal and GPU rendering. [1 lecture]
- **Global Illumination.** Realistic global lighting techniques. [1 lecture]
- **Virtual Reality.** Technology and best practices for an emerging medium. [1 lecture]

Objectives

On completing the course, students should be able to

- use graphics hardware to render interactive images, both polygonal and implicit;
- understand the core technologies of ray tracing, rendering, and implicit surfaces;
- learn techniques of computational geometry and their applications to visualization;
- describe the underlying theory of splines and subdivision and define the Catmull-Clark and Doo-Sabin subdivision methods;
- understand several global illumination technologies such as radiosity and ambient occlusion;
Recommended reading

Students should expect to refer to one or more of these books, but should not find it necessary to purchase any of them.


Further Java

*Lecturer: Dr A.R. Beresford and Dr A. Rice*

*No. of practical classes: 5 x 2-hour sessions (for which supervisions should be given)*

*Suggested hours of supervisions: 2*

*Prerequisite course: Object-Oriented Programming*

*Companion courses: Concurrent and Distributed Systems*

*This course is a prerequisite for the Group Project.*

Aims

The goal of this course is to provide students with the ability to understand the advanced programming features available in the Java programming language, completing the coverage of the language started in the Programming in Java course. The course is designed to accommodate students with diverse programming backgrounds; consequently Java is taught from first principles in a practical class setting where students can work at their own pace from a course handbook. Each practical class will culminate in an assessed exercise.

Practical classes

- **Communication and client applications.** This class will introduce the Eclipse development environment. Students will write a simple client to send and receive data to a server via TCP.

- **Serialisation, reflection and class loaders.** This class will introduce object serialisation. Students will use a class loader and reflection to inspect an object which is only available at run-time.
• **Concurrency and synchronisation.** This class introduces the concurrency and synchronisation primitives found in Java. Students will implement a thread-safe first-in-first-out queue and learn about Java generics.

• **Server applications.** Students implement a server in Java which is capable of communicating concurrently with multiple clients.

• **Vector clocks.** This week students will use the concept of vector clocks to make their client and server robust to message delays and reordering.

**Objectives**

At the end of the course students should

• understand different mechanisms for communication between distributed applications and be able to evaluate their trade-offs;

• be able to use Java generics and annotations to improve software usability, readability and safety;

• understand and be able to exploit the Java class-loading mechanism;

• understand and be able to use concurrency control correctly;

• be able to implement a vector clock algorithm and the happens-before relation.

**Recommended reading**


[http://docs.oracle.com/javase/specs/jls/se8/html/](http://docs.oracle.com/javase/specs/jls/se8/html/)

**Group Project**

*Lecturer: Professor A.F. Blackwell and Dr R. Mortier*

*No. of lectures: 1*

*Prerequisite courses: Software and Security Engineering, Further Java*

**Aims**

The aim of this course is to give students a realistic introduction to software development as practised in industry. This means working to rigid deadlines, with a team of colleagues not of one’s own choosing, having to satisfy an external client that a design brief has been properly interpreted and implemented, all within the constraints of limited effort and technical resources.
Lectures

- **Initial project briefing.** Software engineering: design, quality and management, application of course material. Introduction to possible design briefs. Formation of groups, selection of tools, review meetings.

- **Administrative arrangements.** Announcement of group members. Deliverables: functional specification and module design, module implementation and testing, system integration, testing and documentation. Timetable. Advice on specific tools. First project meeting.

- **Presentation techniques.** Public speaking techniques and the effective use of audio-visual aids. Planning a talk; designing a presentation; common mistakes to avoid.

Objectives

At the end of the course students should

- have a good understanding of how software is developed;

- have consolidated the theoretical understanding of software development acquired in the Software Design course;

- appreciate the importance of planning and controlling a project, and of documentation and presentation;

- have gained confidence in their ability to develop significant software projects and Part IB students should be prepared for the personal project they will undertake in Part II.

Programming in C and C++

*Lecturer: Dr N. Krishnaswami and Professor A. Mycroft*

*No. of lectures: 12*

*Suggested hours of supervisions: 3*

*Prerequisite courses: None, though Operating Systems would be helpful.*

Aims

The aims of this course are to provide a solid introduction to programming in C and to provide an overview of the principles and constraints that affect the way in which the C programming language has been designed and is used, including the differences between it and C++. 
Lectures

- **Introduction to the C language.** Background and goals of C. Types and variables. Expressions and statements. Functions. Multiple compilation units. Tooling for C programming. [2 lectures]

- **Further C concepts.** Preprocessor. Pointers and pointer arithmetic. Data structures. Dynamic memory management. Examples. [2 lectures]


- **Memory Hierarchy and Cache Optimization** Cache hierarchy. Data structure layouts. Intrusive lists. Array-of-structs vs struct-of-array representations. [1 lecture]

- **Linkers, loaders and debugging.** Executable sections. Debug symbols. Inspecting program state. [1 lecture]

- **C semantics.** Undefined vs implementation-defined behaviour. Common optimisation problems. Buffer and integer overflows. Examples. [1 lecture]

- **Introduction to C++.** Goals of C++. Differences between C and C++. References versus pointers. Overloading functions. [1 lecture]


Objectives

At the end of the course students should

- be able to read and write C programs;
- understand the interaction between C programs and the host operating system;
- be familiar with the structure of C program execution in machine memory;
- understand the potential dangers of writing programs in C;
- understand the main differences between C and C++.

Recommended reading

Semantics of Programming Languages

Lecturer: Professor P. Sewell

No. of lectures: 12

Suggested hours of supervisions: 3

This course is a prerequisite for the Part II courses Topics in Concurrency, Hoare Logic and Model Checking and Types.

Aims

The aim of this course is to introduce the structural, operational approach to programming language semantics. It will show how to specify the meaning of typical programming language constructs, in the context of language design, and how to reason formally about semantic properties of programs.

Lectures

- **Introduction.** Transition systems. The idea of structural operational semantics. Transition semantics of a simple imperative language. Language design options. [2 lectures]

- **Types.** Introduction to formal type systems. Typing for the simple imperative language. Statements of desirable properties. [2 lectures]


- **Functions.** Call-by-name and call-by-value function application, semantics and typing. Local recursive definitions. [2 lectures]

- **Data.** Semantics and typing for products, sums, records, references. [1 lecture]

- **Subtyping.** Record subtyping and simple object encoding. [1 lecture]

- **Semantic equivalence.** Semantic equivalence of phrases in a simple imperative language, including the congruence property. Examples of equivalence and non-equivalence. [1 lecture]

- **Concurrency.** Shared variable interleaving. Semantics for simple mutexes; a serializability property. [1 lecture]

Objectives

At the end of the course students should

- be familiar with rule-based presentations of the operational semantics and type systems for some simple imperative, functional and interactive program constructs;
• be able to prove properties of an operational semantics using various forms of induction (mathematical, structural, and rule-based);

• be familiar with some operationally-based notions of semantic equivalence of program phrases and their basic properties.

**Recommended reading**


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**Unix Tools**

*Lecturer: Dr M.G. Kuhn*

*No. of lectures: 8*

*Suggested hours of supervisions: 0–1* (non-examinable course with exercises)

*Prerequisite courses: Operating Systems.*

*This course is a recommended prerequisite for Security.*

**Aims**

This course gives students who have already basic Unix/Linux experience some additional practical software-engineering knowledge: how to use the shell and related tools as an efficient working environment, how to automate routine tasks, and how version control and automated-build tools can help to avoid confusion and accidents, especially when working in teams. These are essential skills, both in industrial software development and student projects.

**Lectures**

- **Unix concepts.** Brief review of Unix history and design philosophy, documentation, terminals, inter-process communication mechanisms and conventions, shell, command-line arguments, environment variables, file descriptors.

- **Shell concepts.** Program invocation, redirection, pipes, file-system navigation, argument expansion, quoting, job control, signals, process groups, variables, locale, history and alias functions, security considerations.

- **Scripting.** Plain-text formats, executables, #!, shell control structures and functions. Startup scripts.
• **Text, file and networking tools.** sed, grep, chmod, find, ssh, rsync, tar, zip, etc.

• **Revision control systems.** diff, patch, RCS, Subversion, git.

• **Software development tools.** C compiler, linker, debugger, make.

• **Perl.** Introduction to a powerful scripting and text-manipulation language. [2 lectures]

**Objectives**

At the end of the course students should

• be confident in performing routine user tasks on a POSIX system, understand command-line user-interface conventions and know how to find more detailed documentation;

• appreciate how simple tools can be combined to perform a large variety of tasks;

• be familiar with the most common tools, file formats and configuration practices;

• be able to understand, write, and maintain shell scripts and makefiles;

• appreciate how using revision control systems and fully automated build processes help to maintain reproducibility and audit trails during software development;

• know enough about basic development tools to be able to install, modify and debug C source code;

• have understood the main concepts of, and gained initial experience in, writing Perl scripts (excluding the facilities for object-oriented programming).

**Recommended reading**

Lent Term 2020: Part IB lectures

Compiler Construction

Lecturer: Dr T.G. Griffin

No. of lectures: 16

Suggested hours of supervisions: 4

Prerequisite: Discrete Mathematics (Part IA)

This course is a prerequisite for Optimising Compilers (Part II).

Aims

This course aims to cover the main concepts associated with implementing compilers for programming languages. We use a running example called SLANG (a Small LANGuage) inspired by the languages described in 1B Semantics. A toy compiler (written in ML) is provided, and students are encouraged to extend it in various ways.

Lectures

- **Overview of compiler structure** The spectrum of interpreters and compilers; compile-time and run-time. Compilation as a sequence of translations from higher-level to lower-level intermediate languages, where each translation preserves semantics. The structure of a simple compiler: lexical analysis and syntax analysis, type checking, intermediate representations, optimisations, code generation. Overview of run-time data structures: stack and heap. Virtual machines. [1 lecture]


- **Compiler Correctness** Recursive functions can be transformed into iterative functions using the Continuation-Passing Style (CPS) transformation. CPS applied to a (recursive) SLANG interpreter to derive, in a step-by-step manner, a correct stack-based compiler. [3 lectures]

- **Data structures, procedures/functions** Representing tuples, arrays, references. Procedures and functions: calling conventions, nested structure, non-local variables. Functions as first-class values represented as closures. Simple optimisations: inline expansion, constant folding, elimination of tail recursion, peephole optimisation. [5 lectures]

- **Advanced topics** Run-time memory management (garbage collection). Static and dynamic linking. Objects and inheritance; implementation of method dispatch. Try-catch exception mechanisms. Compiling a compiler via bootstrapping. [4 lectures]
Objectives

At the end of the course students should understand the overall structure of a compiler, and will know significant details of a number of important techniques commonly used. They will be aware of the way in which language features raise challenges for compiler builders.

Recommended reading

http://www.diku.dk/~torbenm/Basics.

Computation Theory

*Lecturer: Professor A.M. Pitts*

*No. of lectures:* 12

*Suggested hours of supervisions:* 3

*Prerequisite course: Discrete Mathematics*

This course is a prerequisite for Complexity Theory (Part IB).

Aims

The aim of this course is to introduce several apparently different formalisations of the informal notion of algorithm; to show that they are equivalent; and to use them to demonstrate that there are uncomputable functions and algorithmically undecidable problems.

Lectures

- **Introduction: algorithmically undecidable problems.** Decision problems. The informal notion of algorithm, or effective procedure. Examples of algorithmically undecidable problems. [1 lecture]

- **Register machines.** Definition and examples; graphical notation. Register machine computable functions. Doing arithmetic with register machines. [1 lecture]

- **Universal register machine.** Natural number encoding of pairs and lists. Coding register machine programs as numbers. Specification and implementation of a universal register machine. [2 lectures]
• **Undecidability of the halting problem.** Statement and proof. Example of an uncomputable partial function. Decidable sets of numbers; examples of undecidable sets of numbers. [1 lecture]

• **Turing machines.** Informal description. Definition and examples. Turing computable functions. Equivalence of register machine computability and Turing computability. The Church-Turing Thesis. [2 lectures]

• **Primitive and partial recursive functions.** Definition and examples. Existence of a recursive, but not primitive recursive function. A partial function is partial recursive if and only if it is computable. [2 lectures]

• **Lambda-Calculus.** Alpha and beta conversion. Normalization. Encoding data. Writing recursive functions in the lambda-calculus. The relationship between computable functions and lambda-definable functions. [3 lectures]

**Objectives**

At the end of the course students should

• be familiar with the register machine, Turing machine and lambda-calculus models of computability;

• understand the notion of coding programs as data, and of a universal machine;

• be able to use diagonalisation to prove the undecidability of the Halting Problem;

• understand the mathematical notion of partial recursive function and its relationship to computability.

**Recommended reading**

* Hopcroft, J.E., Motwani, R. & Ullman, J.D. (2001). *Introduction to automata theory, languages, and computation*. Addison-Wesley (2nd ed.).


Computer Networking

Lecturer: Dr A.W. Moore

No. of lectures: 20

Suggested hours of supervisions: 5

This course is a prerequisite for the Part II course Principles of Communication

Aims

The aim of this course is to introduce key concepts and principles of computer networks. The course will use a top-down approach to study the Internet and its protocol stack. Instances of architecture, protocol, application-examples will include email, web and media-streaming. We will cover communications services (e.g., TCP/IP) required to support such network applications. The implementation and deployment of communications services in practical networks: including wired and wireless LAN environments, will be followed by a discussion of issues of network-management. Throughout the course, the Internet’s architecture and protocols will be used as the primary examples to illustrate the fundamental principles of computer networking.

Lectures

- **Introduction.** Overview of networking using the Internet as an example. LANs and WANs. OSI reference model, Internet TCP/IP Protocol Stack. Circuit-switching, packet-switching, Internet structure, networking delays and packet loss. [3 lectures]

- **Link layer and local area networks.** Link layer services, error detection and correction, Multiple Access Protocols, link layer addressing, Ethernet, hubs and switches, Point-to-Point Protocol. [2 lectures]

- **Wireless and mobile networks.** Wireless links and network characteristics, Wi-Fi: IEEE 802.11 wireless LANs. [1 lecture]

- **Network layer addressing.** Network layer services, IP, IP addressing, IPv4, DHCP, NAT, ICMP, IPv6. [3 lectures]

- **Network layer routing.** Routing and forwarding, routing algorithms, routing in the Internet, multicast. [3 lectures]

- **Transport layer.** Service models, multiplexing/demultiplexing, connection-less transport (UDP), principles of reliable data transfer, connection-oriented transport (TCP), TCP congestion control, TCP variants. [6 lectures]

- **Application layer.** Client/server paradigm, WWW, HTTP, Domain Name System, P2P. [1.5 lectures]

- **Multimedia networking.** Networked multimedia applications, multimedia delivery requirements, multimedia protocols (SIP), content distribution networks. [0.5 lecture]
Objectives

At the end of the course students should

- be able to analyse a communication system by separating out the different functions provided by the network;
- understand that there are fundamental limits to any communications system;
- understand the general principles behind multiplexing, addressing, routing, reliable transmission and other stateful protocols as well as specific examples of each;
- understand what FEC is;
- be able to compare communications systems in how they solve similar problems;
- have an informed view of both the internal workings of the Internet and of a number of common Internet applications and protocols.

Recommended reading


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Paper 7: Further Human–Computer Interaction

This course is only taken by Part IB and Part II Paper 7 students.

*Lecturer: Professor A.F. Blackwell and Luke Church*

*No. of lectures: 8*

*Suggested hours of supervisions: 2*

*Prerequisite courses: Interaction Design (Part IA)*

Aims

This aim of this course is to provide an introduction to the theoretical foundations of Human Computer Interaction, and an understanding of how these can be applied to the design of complex technologies.
Lectures

- **Theory driven approaches to HCI.** What is a theory in HCI? Why take a theory driven approach to HCI?

- **Design of visual displays.** Segmentation and variables of the display plane. Modes of correspondence.

- **Goal-oriented interaction.** Using cognitive theories of planning, learning and understanding to understand user behaviour, and what they find hard.

- **Designing smart systems.** Using statistical methods to anticipate user needs and actions with Bayesian strategies.

- **Designing efficient systems.** Measuring and optimising human performance through quantitative experimental methods.

- **Designing meaningful systems.** Qualitative research methods to understand social context and requirements of user experience.

- **Evaluating interactive system designs.** Approaches to evaluation in systems research and engineering, including Part II Projects.

- **Designing complex systems.** Worked case studies of applying the theories to a hard HCI problem. Research directions in HCI.

Objectives

At the end of the course students should be able to apply theories of human performance and cognition to system design, including selection of appropriate techniques to analyse, observe and improve the usability of a wide range of technologies.

Recommended reading


Further reading:


Logic and Proof

*Lecturer: Professor L.C. Paulson*

*No. of lectures: 12*

*Suggested hours of supervisions: 3*

*This course is a prerequisite for the Part II courses Machine Learning and Bayesian Inference, Hoare Logic & Model Checking and Natural Language Processing.*
Aims

This course will teach logic, especially the predicate calculus. It will present the basic principles and definitions, then describe a variety of different formalisms and algorithms that can be used to solve problems in logic. Putting logic into the context of Computer Science, the course will show how the programming language Prolog arises from the automatic proof method known as resolution. It will introduce topics that are important in mechanical verification, such as binary decision diagrams (BDDs), SAT solvers and modal logic.

Lectures

- **Introduction to logic.** Schematic statements. Interpretations and validity. Logical consequence. Inference.

- **Propositional logic.** Basic syntax and semantics. Equivalences. Normal forms. Tautology checking using CNF.

- **The sequent calculus.** A simple (Hilbert-style) proof system. Natural deduction systems. Sequent calculus rules. Sample proofs.

- **First order logic.** Basic syntax. Quantifiers. Semantics (truth definition).


- **Satisfiability Modulo Theories.** Decision problems and procedures. How SMT solvers work.


- **Tableaux methods.** Simplifying the sequent calculus. Examples. Adding unification. Skolemisation. The world's smallest theorem prover?
Objectives

At the end of the course students should

- be able to manipulate logical formulas accurately;
- be able to perform proofs using the presented formal calculi;
- be able to construct a small BDD;
- understand the relationships among the various calculi, e.g. SAT solving, resolution and Prolog;
- understand the concept of a decision procedure and the basic principles of “satisfiability modulo theories”.
- be able to apply the unification algorithm and to describe its uses.

Recommended reading


Paper 7: Prolog

This course is only taken by Part IB and Part II Paper 7 students.

* Lecturer: Dr. A. Rice
* No. of lectures: 8
* Suggested hours of supervisions: 2
* Prerequisite courses: Foundations of Computer Science, Algorithms

Aims

The aim of this course is to introduce programming in the Prolog language. Prolog encourages a different programming style to Java or ML and particular focus is placed on programming to solve real problems that are suited to this style. Practical experimentation with the language is strongly encouraged.

Lectures

- **Introduction to Prolog**. The structure of a Prolog program and how to use the Prolog interpreter. Unification. Some simple programs.
• **Arithmetic and lists.** Prolog’s support for evaluating arithmetic expressions and lists. The space complexity of program evaluation discussed with reference to last-call optimisation.

• **Backtracking, cut, and negation.** The cut operator for controlling backtracking. Negation as failure and its uses.

• **Search and cut.** Prolog’s search method for solving problems. Graph searching exploiting Prolog’s built-in search mechanisms.

• **Difference structures.** Difference lists: introduction and application to example programs.

• **Building on Prolog.** How particular limitations of Prolog programs can be addressed by techniques such as Constraint Logic Programming (CLP) and tabled resolution.

**Objectives**

At the end of the course students should

• be able to write programs in Prolog using techniques such as accumulators and difference structures;

• know how to model the backtracking behaviour of program execution;

• appreciate the unique perspective Prolog gives to problem solving and algorithm design;

• understand how larger programs can be created using the basic programming techniques used in this course.

**Recommended reading**

* Bratko, I. (2001). *PROLOG programming for artificial intelligence*. Addison-Wesley (3rd or 4th ed.).

Further reading:

O’Keefe, R. (1990). *The craft of Prolog*. MIT Press. [This book is beyond the scope of this course, but it is very instructive. If you understand its contents, you’re more than prepared for the examination.]
Easter Term 2020: Part IB lectures

Artificial Intelligence

Lecturer: Dr S.B. Holden

No. of lectures: 12

Suggested hours of supervisions: 3

Prerequisite courses: Algorithms. In addition the course requires some mathematics, in particular some use of vectors and some calculus. Part IA Natural Sciences Mathematics or equivalent and Discrete Mathematics are likely to be helpful although not essential. Similarly, elements of Foundations of Data Science, Logic and Proof, Prolog and Complexity Theory are likely to be useful.

This course is a prerequisite for the Part II courses Machine Learning and Bayesian Inference and Natural Language Processing.

Aims

The aim of this course is to provide an introduction to some fundamental issues and algorithms in artificial intelligence (AI). The course approaches AI from an algorithmic, computer science-centric perspective; relatively little reference is made to the complementary perspectives developed within psychology, neuroscience or elsewhere. The course aims to provide some fundamental tools and algorithms required to produce AI systems able to exhibit limited human-like abilities, particularly in the form of problem solving by search, game-playing, representing and reasoning with knowledge, planning, and learning.

Lectures

- **Introduction.** Alternate ways of thinking about AI. Agents as a unifying view of AI systems. [1 lecture]

- **Search I.** Search as a fundamental paradigm for intelligent problem-solving. Simple, uninformed search algorithms. Tree search and graph search. [1 lecture]

- **Search II.** More sophisticated heuristic search algorithms. The A* algorithm and its properties. Improving memory efficiency: the IDA* and recursive best first search algorithms. Local search and gradient descent. [1 lecture]

- **Game-playing.** Search in an adversarial environment. The minimax algorithm and its shortcomings. Improving minimax using alpha-beta pruning. [1 lecture]

- **Constraint satisfaction problems (CSPs).** Standardising search problems to a common format. The backtracking algorithm for CSPs. Heuristics for improving the search for a solution. Forward checking, constraint propagation and arc consistency. [1 lecture]
• **Backjumping in CSPs.** Backtracking, backjumping using Gaschnig’s algorithm, graph-based backjumping. [1 lecture]

• **Knowledge representation and reasoning I.** How can we represent and deal with commonsense knowledge and other forms of knowledge? Semantic networks, frames and rules. How can we use inference in conjunction with a knowledge representation scheme to perform reasoning about the world and thereby to solve problems? Inheritance, forward and backward chaining. [1 lectures]

• **Knowledge representation and reasoning II.** Knowledge representation and reasoning using first order logic. The frame, qualification and ramification problems. The situation calculus. [1 lectures]

• **Planning I.** Methods for planning in advance how to solve a problem. The STRIPS language. Achieving preconditions, backtracking and fixing threats by promotion or demotion: the partial-order planning algorithm. [1 lecture]


• **Neural Networks I.** A brief introduction to supervised learning from examples. Learning as fitting a curve to data. The perceptron. Learning by gradient descent. [1 lecture]

• **Neural Networks II.** Multilayer perceptrons and the backpropagation algorithm. [1 lecture]

**Objectives**

At the end of the course students should:

• appreciate the distinction between the popular view of the field and the actual research results;

• appreciate the fact that the computational complexity of most AI problems requires us regularly to deal with approximate techniques;

• be able to design basic problem solving methods based on AI-based search, knowledge representation, reasoning, planning, and learning algorithms.

**Recommended reading**

The recommended text is:


There are many good books available on artificial intelligence; one alternative is:

For some of the material you might find it useful to consult more specialised texts, in particular:


Complexity Theory

Lecturer: Professor A. Dawar

No. of lectures: 12

Suggested hours of supervisions: 3

Prerequisite courses: Algorithms, Computation Theory

Aims

The aim of the course is to introduce the theory of computational complexity. The course will explain measures of the complexity of problems and of algorithms, based on time and space used on abstract models. Important complexity classes will be defined, and the notion of completeness established through a thorough study of NP-completeness. Applications to cryptography will be considered.

Lectures

• **Algorithms and problems.** Complexity of algorithms and of problems. Lower and upper bounds. Examples: sorting and travelling salesman.

• **Time and space.** Models of computation and measures of complexity. Time and space complexity on a Turing machine. Decidability and complexity.

• **Time complexity.** Time complexity classes. Polynomial time problems and algorithms. Problems on numbers, graphs and formulas.

• **Non-determinism.** Non-deterministic machines. The complexity class NP and its various characterizations. Non-deterministic algorithms for satisfiability and other problems in NP.

• **NP-completeness.** Reductions and completeness. NP-completeness of satisfiability.

• **More NP-complete problems.** Graph-theoretic problems. Independent set, clique and 3-colourability.
• **More NP-complete problems.** Sets, numbers and scheduling. Matching, set covering and knapsack.

• **coNP.** Validity of boolean formulae and its completeness. NP \( \cap \) coNP. Primality and factorisation.

• **Cryptographic complexity.** One-way functions. The class UP.

• **Space complexity.** Deterministic and non-deterministic space complexity classes. The reachability method. Savitch’s theorem.

• **Hierarchy.** The time and space hierarchy theorems and complete problems.

• **Descriptive complexity.** Logics capturing complexity classes. Fagin’s theorem.

**Objectives**

At the end of the course students should

• be able to analyse practical problems and classify them according to their complexity;

• be familiar with the phenomenon of NP-completeness, and be able to identify problems that are NP-complete;

• be aware of a variety of complexity classes and their interrelationships;

• understand the role of complexity analysis in cryptography.

**Recommended reading**


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**Paper 7: Concepts in Programming Languages**

This course is only taken by Part IB and Part II Paper 7 students.

*Lecturer: Professor A. Mycroft*

*No. of lectures: 8*

*Suggested hours of supervisions: 2*

*Prerequisite courses: None.*
Aims

The general aim of this course is to provide an overview of the basic concepts that appear in modern programming languages, the principles that underlie the design of programming languages, and their interaction.

Lectures


- **The ancestors: Fortran, Lisp, Algol and Pascal.** Key ideas: procedural (Fortran), declarative (Lisp), block structured (Algol and Pascal). Execution models (abstract machines), data types, control structures, storage, arrays and pointers, procedures and forms of parameter passing, scope, strict and lazy evaluation, garbage collection. Programs as data (Lisp).


- **Languages for parallel processing.** Shared-memory concurrency with spawn/sync (OpenMP, Cilk, X10). Distributed-memory models (the actor model, Erlang). External vs. internal iteration.

- **Types.** Types in programming languages. Type safety. Type systems—static vs. dynamic. Type checking and type inference. Polymorphism. Overloading. Type equivalence.


- **More-advanced concepts and idioms.** Haskell monads, type classes. Continuation passing style and call/cc. Dependent types.

Objectives

At the end of the course students should

- be familiar with several language paradigms and how they relate to different application domains;
- understand the design space of programming languages, including concepts and constructs from past languages as well as those that may be used in the future;
• develop a critical understanding of the programming languages that we use by being able to identify and compare the same concept as it appears in different languages.

Recommended reading

Books:

Papers:

Paper 7: Formal Models of Language

This course is only taken by Part IB and Part II Paper 7 students.

Lecturers: Dr P. Buttery

No. of lectures: 8

Suggested hours of supervisions: 2

Prerequisite courses: Discrete Maths; Compiler Construction;

Aims

This course studies formal models of language and considers how they might be relevant to the processing and acquisition of natural languages. The course will extend knowledge of formal language theory; introduce several new grammars; and use concepts from information theory to describe natural language.
Lectures

- **Natural language and the Chomsky hierarchy 1.** Recap classes of language. Closure properties of language classes. Recap pumping lemma for regular languages. Discussion of relevance (or not) to natural languages (example embedded clauses in English).

- **Natural language and the Chomsky hierarchy 2.** Pumping lemma for context free languages. Discussion of relevance (or not) to natural languages (example Swiss-German cross serial dependancies). Properties of minimally context sensitive languages. Introduction to tree adjoining grammars.

- **Language processing and context free grammar parsing 1.** Recap of context free grammar parsing. Language processing predictions based on top down parsing models (example Yngve's language processing predictions). Language processing predictions based on probabilistic parsing (example Halle's language processing predictions).

- **Language processing and context free grammar parsing 2.** Introduction to context free grammar equivalent dependancy grammars. Language processing predictions based on Shift-Reduce parsing (examples prosodic look-ahead parsers, Parsey McParseface).

- **Grammar induction of language classes.** Introduction to grammar induction. Discussion of relevance (or not) to natural language acquisition. Gold's theorem. Introduction to context free grammar equivalent categorial grammars and their learnable classes.

- **Natural language and information theory 1.** Entropy and natural language typology. Uniform information density as a predictor for language processing.

- **Natural language and information theory 2.** Noisy channel encoding as a model for spelling error, translation and language processing.

- **Vector space models and word vectors.** Introduction to word vectors (example Word2Vec). Word vectors as predictors for semantic language processing.

Objectives

At the end of the course students should

- understand how known natural languages relate to formal languages in the Chomsky hierarchy;
- have knowledge of several context free grammars equivalents;
- understand how we might make predictions about language processing and language acquisition from formal models;
- know how to use information theoretic concepts to describe aspects of natural language.
Recommended reading

* Jurafsky, D. & Martin, J. (2008). *Speech and language processing*. Prentice Hall.

Security

*Lecturer: Dr M.G. Kuhn*

*No. of lectures: 12*

*Suggested hours of supervisions: 3*

*Prerequisite courses: Operating Systems; Computer Networking; Programming in C; Unix Tools (recommended)*

**Aims**

This course provides an overview of technical measures commonly used to enforce security policies, to protect networked and multi-user information systems against malicious user activity, mainly at the level of operating systems and network protocols. It also discusses common security concepts and pitfalls for application programmers and system architects, and strategies for exploiting and mitigating the resulting vulnerabilities.

**Lectures**


- **Operating-system security overview.** Access-control matrix, trusted computing base, domain separation, CPU modes, system calls, residual information protection, virtual machines. [0.5 lecture]

- **POSIX discretionary access control.** User and group databases and identifiers, file permission modes, ownership rights, sticky bit, group inheritance, set-uid, elevation of privileges, root user, NFS root squash, chroot, POSIX.1e ACLs.

- **Windows discretionary access control.** NTFS access rights, security identifiers, access-control entries and lists, inheritance, services, auditing, NFSv4 ACLs.

- **Linux-specific mechanisms.** PAM, LSM, Linux capabilities, AppArmor, seccomp, eBPF, audit, cgroups, namespaces, containers.
• **Running untrusted code.** Mandatory access control, covert channels, SELinux, type enforcement, iOS/macOS/Android app-store sandboxes, capabilities.

• **Software vulnerabilities.** Buffer/integer overflows, ASLR, metacharacter vulnerabilities: shell and SQL injection, side channels, race conditions, environmental exploits, fuzzing.

• **Cryptography overview.** Private/public-key encryption, MACs, digital signatures, certificates, key revocation, secure hash functions, key-establishment schemes, key generation. [0.5 lecture]

• **Entity authentication.** Password verification, guessing user-generated secrets, biometric identification, hardware tokens, challenge-response authentication protocols, Kerberos, ssh, TLS.

• **Network access.** Ethernet hubs and switches, ARP/NDP/DHCP spoofing, 802.1q VLAN tagging/trunking/hopping, port isolation, 802.1x, RADIUS, EAP, Wifi, GSM/LTE.

• **Internet protocols.** TCP vs UDP, firewalls, iptables, IPSEC/IKE, VPNs, IP options/fragmentation, DDoS.

• **Email and DNS security.** SMTP/RFC822 header forgery, spam, SPF, DKIM, DNS vulnerabilities, DNSSEC.

• **Web security.** HTTP basics, HTTPS, SNI, HTTP authentication, cookies, single sign-on (Ucam WebAuth, SAML), delegation (OAuth2), JavaScript, cross-site scripting, cross-site request forgery, same-origin policy, CORS.

**Objectives**

By the end of the course, students should appreciate the importance of adversarial thinking in systems design and have a good overview of the security mechanisms and attributes of some of the most commonly used operating systems, networking infrastructure and Internet applications. They should also understand commonly exploited vulnerabilities of authentication mechanisms and know how to avoid some common security pitfalls in software development.

**Recommended reading**

Introduction to Part II

This document lists the courses offered by the Computer Laboratory for Part II of the Computer Science Tripos. Separate booklets give details of the syllabus for other Parts of the Computer Science Tripos.

Students taking the Part II 50% option of the CST will read papers 7, 8 and 9 and submit a dissertation. Each of these four is marked out of 100 giving a total available credit in Part II of 400 marks. Alternatively, those taking the Part II 75% option will take papers 8, 9, submit a dissertation, and offer two units of assessment.

Some courses are specific to either Paper 7 or are Units of Assessment, and have been marked as such in this booklet. Those students following the 75% Computer Science option, who have taken Paper 7 in Part II, will additionally choose two of the units of assessment. Those students taking the 50% Computer Science option, will attend the Paper 7 courses instead. The remaining courses can be chosen by all Part II students.

The taught modules in Part II are examined in papers 7, 8 and 9 and you answer five questions from each paper. There are no restrictions on which questions you answer. The layout of the papers is announced just before the Michaelmas term starts, but it is generally mostly the same as in previous years, varying only to accommodate new, withdrawn or suspended courses. The units are assessed in a variety of ways, and details can be found in the course descriptions.

It is up to you to make sure you read sufficient courses to be able to answer five questions on each of the papers. Generally, you should aim to be able to answer at least six questions on each paper. You are certainly not expected to go to all the Part II lectures and be able to answer all of the questions on every paper — that would be more or less impossible.

Here is a suggestion for how to plan your courses: In September, just before the start of the year, look through the course list and strike out any course you know you won’t do (i.e. remove the definite ‘no’s - there are always some). Then attend the first lecture of every Part II course to get the feel for it and make a decision on whether to continue after checking that dropping the course doesn’t leave you short on any paper. Work on the basis of being able to answer 6 questions, with a 7th as a backup where you are confident of scoring half marks (but probably no more).

It is the duty of your Director of Studies to advise you in course selection so do ask for guidance.

The syllabus information given here is for guidance only and should not be considered definitive. Current timetables can be found at [http://www.cl.cam.ac.uk/teaching/timetables/](http://www.cl.cam.ac.uk/teaching/timetables/)

For most of the courses listed below, a list of recommended books is given. These are roughly in order of usefulness, and lecturers have indicated by means of an asterisk those books which are most recommended for purchase by College libraries.

The Computer Laboratory Library aims to keep at least one copy of each of the course texts in "The Booklocker" (see [http://www.cl.cam.ac.uk/library/](http://www.cl.cam.ac.uk/library/)).
For copies of the other syllabus booklets and for answers to general enquiries about Computer Science courses, please get in touch with:

Teaching Administrator
University of Cambridge
Computer Laboratory
William Gates Building
J J Thomson Avenue
Cambridge
CB3 0FD

telephone: 01223 763505
fax: 01223 334678
e-mail: teaching-admin@cl.cam.ac.uk
Michaelmas Term 2019: Part II lectures

Unit: Advanced Graphics and Image Processing

This course is only taken by Part II 75% students.

Lecturers: Dr R. Mantiuk

No. of lectures and practical classes: 14

Prerequisite courses: Programming in C.

Capacity: 30-50

Aims

Advanced Graphics covers topics related to processing, perception and display of images. The focus of the course is on the algorithms behind new emerging display technologies, such as virtual reality, augmented reality, and high dynamic range displays. It complements two computer graphics courses, Introduction to Graphics and Further Graphics, by introducing problems that became the part of graphics pipeline: tone-mapping, post-processing, displays and models of visual perception.

Lectures

- **GP-GPU**: scheduling and thread mapping, reductions.
- **Advanced image processing**: edge-stopping filters, pyramids, optimization-based image processing.
- **Beyond 2D**: stereo rendering and light fields.
- **Models of visual perception**: visual system, brightness perception, detection and discrimination, contrast sensitivity function, contrast constancy, perceptually uniform spaces, depth perception.
- **High Dynamic Range and tone mapping**: dynamic range, display model, methods of tone-mapping.
- **Display technologies**: 2D displays, 3D displays, temporal display characteristic, HDR displays.
- **Virtual and Augmented Reality**: display technologies, VR rendering, orientation tracking, pose tracking, perceptual considerations, panoramic imaging.

Objectives

By the end of the course students should be able to:

- implement real-time image processing methods on a GPU (OpenCL);
• design and implement a tone-mapping algorithm;

• describe the limitations of display technologies (dynamic range, brightness, visual comfort, VR simulation sickness) and how they can be addressed using computational methods (tone-mapping, HDR displays);

• describe the limitations of the visual system and how those limitation can be exploited in computer graphics and image processing.

Recommended reading

Reinhard, E., et. al. (2010) High Dynamic Range Imaging: Acquisition, Display, and Image-Based Lighting. Morgan Kaufmann (2nd ed.).

Bioinformatics

Lecturer: Professor P. Lio'
No. of lectures: 12
Suggested hours of supervisions: 3

Aims

This course focuses on algorithms used in Bioinformatics and System Biology. Most of the algorithms are general and can be applied in other fields on multidimensional and noisy data. All the necessary biological terms and concepts useful for the course and the examination will be given in the lectures. The most important software implementing the described algorithms will be demonstrated.

Lectures

• Introduction to biological data: Bioinformatics as an interesting field in computer science. Computing and storing information with DNA (including Adleman’s experiment).

• Dynamic programming. Longest common subsequence, DNA global and local alignment, linear space alignment, Nussinov algorithm for RNA, heuristics for multiple alignment. (Vol. 1, chapter 5)

• Sequence database search. Blast. (see notes and textbooks)

• Genome sequencing. De Bruijn graph. (Vol. 1, chapter 3)
- **Phylogeny.** Distance based algorithms (UPGMA, Neighbour-Joining). Parsimony-based algorithms. Examples in Computer Science. (Vol. 2, chapter 7)

- **Clustering.** Hard and soft K-means clustering, use of Expectation Maximization in clustering, Hierarchical clustering, Markov clustering algorithm. (Vol. 2, chapter 8)

- **Genomics Pattern Matching.** Suffix Tree String Compression and the Burrows-Wheeler Transform. (Vol. 2, chapter 9)

- **Hidden Markov Models.** The Viterbi algorithm, profile HMMs for sequence alignment, classifying proteins with profile HMMs, soft decoding problem, Baum-Welch learning. (Vol. 2, chapter 10)

**Objectives**

At the end of this course students should

- understand Bioinformatics terminology;
- have mastered the most important algorithms in the field;
- be able to work with bioinformaticians and biologists;
- be able to find data and literature in repositories.

**Recommended reading**


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**Business Studies**

*Lecturer: Jack Lang and Stewart McTavish*

*No. of lectures: 8*

*Suggested hours of supervisions: 2*

*Prerequisite course: Economics, Law & Ethics*

*This course is a prerequisite for E-Commerce.*
Aims

How to start and run a computer company; the aims of this course are to introduce students to all the things that go to making a successful project or product other than just the programming. The course will survey some of the issues that students are likely to encounter in the world of commerce and that need to be considered when setting up a new computer company.

See also Business Seminars in the Easter Term.

Lectures

- **So you’ve got an idea?** Introduction. Why are you doing it and what is it? Types of company. Market analysis. The business plan.


- **Project planning and management.** Role of a manager. PERT and GANTT charts, and critical path analysis. Estimation techniques. Monitoring.


- **Marketing and selling.** Sales and marketing are different. Marketing; channels; marketing communications. Stages in selling. Control and commissions.


Objectives

At the end of the course students should

- be able to write and analyse a business plan;

- know how to construct PERT and GANTT diagrams and perform critical path analysis;

- appreciate the differences between profitability and cash flow, and have some notion of budget estimation;
• have an outline view of company formation, share structure, capital raising, growth and exit routes;

• have been introduced to concepts of team formation and management;

• know about quality documentation and productization processes;

• understand the rudiments of marketing and the sales process.

Recommended reading

Lang, J. (2001). The high-tech entrepreneur’s handbook: how to start and run a high-tech company. FT.COM/Prentice Hall.

Students will be expected to be able to use Microsoft Excel and Microsoft Project.

For additional reading on a lecture-by-lecture basis, please see the course website.

Students are strongly recommended to enter the CU Entrepreneurs Business Ideas Competition [http://www.cue.org.uk/]

Unit: Data Science: principles and practice

This course is only taken by Part II 75% students.

Lecturers: Dr M. Rei, Dr E. Kochmar, Dr D. Wischik and Professor E.J. Briscoe

No. of lectures and practical classes: 12

Prerequisite courses: NST Mathematics, Machine Learning and Real-World Data and Foundations of Data Science.

Capacity: 40-50

Aims

The course will develop core areas of Data Science (eg. models for regression and classification) from several perspectives: conceptual formulation and properties, solution algorithms and their implementation, data visualization for exploratory data analysis and the effective presentation of modelling outputs. The lectures will be complemented by practical classes using Python, scikit-learn and TensorFlow.

Lectures

• Introduction. Motivation, applications, examples, loading common data formats, calculating statistics over a dataset, logistics and overview of the course.

• Linear Regression. Defining a model, fitting a model, least squares regression, linear regression, gradient descent, scikit-learn.
• Practical: Linear Regression

• Classification. Classification, perceptron, logistic regression, multi-class classification, regularisation, kernels, exploratory data visualisation.

• Practical: Classification

• Deep Learning, part I. Training neural networks, applications, multilayer perceptrons, stochastic gradient descent, backpropagation.

• Deep Learning, part II. Advanced architectures, convnets, RNNs, introduction to TensorFlow.

• Practical: Deep Learning

• Visualization, part I. Scales and coordinates, depicting comparisons.

• Visualization, part II. Common plotting patterns, including dimension reduction.

• Practical: Visualization

• Challenges in Data Science. Summary of the course, overview of other relevant techniques, ethics and privacy issues, bias in the training data, information about the hand out test.

Objectives

By the end of the course students should be able to:

• demonstrate understanding and practical skills in Data Science;

• be able to specify and work with an analytical model;

• be able to effectively implement Data Science algorithms;

• understand how data visualization underpins exploring datasets as well as communicating the findings of data science models.

Recommended reading

Python Basic Tutorial. Available online:
https://www.tutorialspoint.com/python/index.htm
Numpy: Quickstart Tutorial. Available online:
https://docs.scipy.org/doc/numpy/user/quickstart.html
Get Started with TensorFlow. Available online:
https://www.tensorflow.org/tutorials/
Denotational Semantics

Lecturer: Professor A.M. Pitts

No. of lectures: 10

Suggested hours of supervisions: 3

Aims

The aims of this course are to introduce domain theory and denotational semantics, and to show how they provide a mathematical basis for reasoning about the behaviour of programming languages.

Lectures

- **Introduction.** The denotational approach to the semantics of programming languages. Recursively defined objects as limits of successive approximations.

- **Least fixed points.** Complete partial orders (cpos) and least elements. Continuous functions and least fixed points.

- **Constructions on domains.** Flat domains. Product domains. Function domains.

- **Scott induction.** Chain-closed and admissible subsets of cpos and domains. Scott’s fixed-point induction principle.


- **Denotational semantics of PCF.** Denotation of types and terms. Compositionality. Soundness with respect to evaluation. [2 lectures].


- **Full abstraction.** Failure of full abstraction for the domain model. PCF with parallel or.

Objectives

At the end of the course students should

- be familiar with basic domain theory: cpos, continuous functions, admissible subsets, least fixed points, basic constructions on domains;

- be able to give denotational semantics to simple programming languages with simple types;
• be able to apply denotational semantics; in particular, to understand the use of least fixed points to model recursive programs and be able to reason about least fixed points and simple recursive programs using fixed point induction;

• understand the issues concerning the relation between denotational and operational semantics, adequacy and full abstraction, especially with respect to the language PCF.

Recommended reading


Unit: Digital Signal Processing

The whole course is for Part II 75% students, while Part II 50% students can take one question in Paper 7 about part 1 of this course.

*Lecturer: Dr M.G. Kuhn*

*No. of lectures: 16*

*Prerequisite courses: Mathematical Methods I and III from the NST Mathematics course (or equivalent), LaTeX and MATLAB (recommended).*

*Aims*

This course teaches the basic signal-processing principles necessary to understand many modern high-tech systems, with application examples focussing on audio processing, audio and image coding, communication systems, and linear feed-back control. Students will gain practical experience from numerical experiments in programming assignments (in MATLAB, NumPy or Julia).

*Lectures*

Part 1 (about 9–10 lectures) focusses on basic theory and audio applications.

• **Signals and systems.** Discrete sequences and systems: types and properties. Amplitude, phase, frequency, modulation, decibels, root-mean square. Linear time-invariant systems, convolution. Some examples from electronics, optics and acoustics.

• **Phasors.** Eigen functions of linear time-invariant systems. Review of complex arithmetic. Phasors as orthogonal base functions.
- **Fourier transform.** Forms and properties of the Fourier transform. Convolution theorem. Rect and sinc.
- **Dirac’s delta function.** Fourier representation of sine waves, impulse combs in the time and frequency domain. Amplitude-modulation in the frequency domain.
- **Discrete sequences and spectra.** Sampling of continuous signals, periodic signals, aliasing, interpolation, sampling and reconstruction, sample-rate conversion, oversampling, spectral inversion.
- **Discrete Fourier transform.** Continuous versus discrete Fourier transform, symmetry, linearity, FFT, real-valued FFT, FFT-based convolution, zero padding, FFT-based resampling, deconvolution exercise.
- **Spectral estimation.** Short-time Fourier transform, leakage and scalloping phenomena, windowing, zero padding. Audio and voice examples. DTFM exercise.
- **Finite impulse-response filters.** Properties of filters, implementation forms, window-based FIR design, use of frequency-inversion to obtain high-pass filters, use of modulation to obtain band-pass filters.
- **Infinite impulse-response filters.** Sequences as polynomials, z-transform, zeros and poles, some analog IIR design techniques (Butterworth, Chebyshev I/II, elliptic filters, second-order cascade form).

Part 2 (about 6–7 lectures) adds material on software-defined radio techniques, statistical signals, audio-visual signal compression and linear feedback control systems.

- **Band-pass signals.** Band-pass sampling and reconstruction, IQ up and down conversion, superheterodyne receivers, software-defined radio front-ends, IQ representation of AM and FM signals and their demodulation.
- **Random sequences and noise.** Random variables, stationary and ergodic processes, autocorrelation, crosscorrelation, deterministic cross-correlation sequences, filtered random sequences, white noise, periodic averaging.
- **Correlation coding.** Entropy, delta coding, linear prediction, dependence versus correlation, random vectors, covariance, decorrelation, matrix diagonalization, eigen decomposition, Karhunen–Loève transform, principal component analysis. Relation to orthogonal transform coding using fixed basis vectors, such as DCT.
- **Lossy versus lossless compression.** What information is discarded by human senses and can be eliminated by encoders? Perceptual scales, audio masking, spatial resolution, colour coordinates, some demonstration experiments.
- **Quantization, image coding standards.** Uniform and logarithmic quantization, A/mu-law coding, dithering, JPEG.
• **Control systems.** Open-loop and closed-loop (feedback) control, PID controllers, stability, phase-locked loops.

**Objectives**

By the end of part 1 of the course students should be able to

• apply basic properties of time-invariant linear systems;
• understand sampling, aliasing, convolution, filtering, the pitfalls of spectral estimation;
• explain the above in time and frequency domain representations;
• use filter-design software;
• visualize and discuss digital filters in the $z$-domain;
• use the FFT for convolution, deconvolution, filtering;
• implement, apply and evaluate simple DSP applications;

By the end of part 2, students should be able to discuss and explain many fundamental concepts of techniques commonly used in digital communication and control systems in terms of the concepts introduced in part 1.

**Recommended reading**


**Information Theory**

*Lecturer: Professor J.G. Daugman*

*No. of lectures: 16*

*In lieu of supervisions, exercises will be set and reviewed in two Examples Classes.*

**Aims**

This course introduces the principles and applications of information theory: how information is measured in terms of probability and various entropies, how these are used to calculate the capacity of communication channels, with or without noise, and to measure how much random variables reveal about each other. Coding schemes including error correcting codes are studied along with data compression, spectral analysis,
transforms, and wavelet coding. Applications of information theory are reviewed, from astrophysics to pattern recognition.

Lectures

- **Foundations: probability, uncertainty, information.** How concepts of randomness, redundancy, compressibility, noise, bandwidth, and uncertainty are related to information. Ensembles, random variables, marginal and conditional probabilities. How the metrics of information are grounded in the rules of probability.

- **Entropies defined, and why they are measures of information.** Marginal joint and conditional entropy; chain rule for entropy. Cross-entropy and distances between distributions. Mutual information between random variables. Why entropy gives fundamental measures of information content.

- **Source coding theorem; prefix, variable-, and fixed-length codes.** Markov sources. Entropy of a multi-state Markov process. Symbol codes; Huffman codes and the prefix property. Binary symmetric channels. Capacity of a noiseless discrete channel.

- **Noisy discrete channel properties, and channel capacity.** Perfect communication through a noisy channel: error-correcting codes. Capacity of a discrete channel as the maximum of its mutual information.

- **Information represented by projections and in transforms.** Expressing data in vector spaces or as a linear combination of basis functions. Inner product spaces and orthonormal systems. Norms, span, and linear subspaces; dimensionality reduction.

- **Fourier analysis: series and transforms, discrete or continuous.** How periodic and aperiodic data are analysed and represented by Fourier methods. Rates of convergence. Information revealed in the Fourier domain. Discrete, inverse, and Fast Fourier Transforms; butterfly algorithm. Duality properties. Wavelet transforms.

- **Spectral properties of continuous-time signals and channels.** Signals represented as combinations of complex exponential eigenfunctions; channels represented as spectral filters that add noise. Convolution. Applying Fourier analysis to communication schemes.

- **Continuous information; density; noisy channel coding theorem.** Extensions of discrete entropies and measures to the continuous case. Signal-to-noise ratio; power spectral density. Gaussian channels. Relative significance of bandwidth and noise limitations. The Shannon rate limit for noisy continuous channels.

- **Signal coding and transmission schemes using Fourier theorems.** Nyquist Sampling Theorem. Aliasing and its prevention. Modulation and shift theorems; multiple carriers; frequency and phase modulation codes; ensembles. Filters, coherence, demodulation; noise removal by correlation.
• The quantized degrees-of-freedom in a continuous signal. Why a continuous signal of finite bandwidth and duration has a fixed number of degrees-of-freedom. Diverse illustrations of the principle that information, even in such a signal, comes in quantized, countable, packets.


• Data compression codes and protocols. Run-length coding; dictionary methods on strings; vector quantisation; JPEG and JP2K image compression; orthogonal subspace projections; predictive coding; the Laplacian pyramid; and wavelet scalar quantisation.

• Kolmogorov complexity. Minimal description length. Definition of the algorithmic complexity of a data sequence, and its relation to the entropy of the distribution from which the data was drawn. Fractals. Minimal description length, and why this measure of complexity is not computable.

• Applications of information theory in other sciences. Use of information metrics and analysis in: genomics; neuroscience; astrophysics; noisy signal classification; and pattern recognition including biometrics.

Objectives

At the end of the course students should be able to

• calculate the information content of a random variable from its probability distribution;

• relate the joint, conditional, and marginal entropies of variables in terms of their coupled probabilities;

• define channel capacities and properties using Shannon’s Theorems;

• construct efficient codes for data on imperfect communication channels;

• generalize the discrete concepts to continuous signals on continuous channels;

• understand encoding and communication schemes in terms of the spectral properties of signals and channels;

• describe compression schemes, and efficient coding using wavelets and other representations for data.

Recommended reading

LaTeX and MATLAB

Lecturer: Dr M.G. Kuhn
No. of lectures: 2

Suggested hours of supervisions: 0–1 (non-examinable course with exercises)

LaTeX skills are useful for preparing the Part II dissertation. MATLAB skills are useful for programming exercises in some Part II courses (e.g. Digital Signal Processing).

Aims

Introduction to two widely-used languages for typesetting dissertations and scientific publications, for prototyping numerical algorithms and to visualize results.

Lectures

- **LaTeX.** Workflow example, syntax, typesetting conventions, non-ASCII characters, document structure, packages, mathematical typesetting, graphics and figures, cross references, build tools.

- **MATLAB.** Tools for technical computing and visualization. The matrix type and its operators, 2D/3D plotting, common functions, function definitions, toolboxes, vectorized audio demonstration.

Objectives

Students should be able to avoid the most common LaTeX mistakes, to prototype simple image and signal processing algorithms in MATLAB, and to visualize the results.

Recommended reading


Mittelbach, F., et al. (2004). *The LaTeX companion*. Addison-Wesley (2nd ed.).

Unit: Multicore Semantics and Programming

This course is only taken by Part II 75% students.

Lecturers: Professor P. Sewell and Dr T. Harris

No. of lectures and practical classes: 8

Prerequisite courses: Discrete mathematics, Object-Oriented programming, Semantics of Programming Languages.
Aims

In recent years multiprocessors have become ubiquitous, but building reliable concurrent systems with good performance remains very challenging. The aim of this module is to introduce some of the theory and the practice of concurrent programming, from hardware memory models and the design of high-level programming languages to the correctness and performance properties of concurrent algorithms.

Lectures

Part 1: Introduction and relaxed-memory concurrency [Peter Sewell]

- **Introduction.** Sequential consistency, atomicity, basic concurrent problems. [1 block]
- **Concurrency on real multiprocessors:** the relaxed memory model(s) for x86, ARM, and IBM Power, and theoretical tools for reasoning about x86-TSO programs. [2 blocks]
- **High-level languages.** An introduction to C/C++11 and Java shared-memory concurrency. [1 block]

Part 2: Concurrent algorithms [Tim Harris]

- **Concurrent programming.** Simple algorithms (readers/writers, stacks, queues) and correctness criteria (linearisability and progress properties). Advanced synchronisation patterns (e.g. some of the following: optimistic and lazy list algorithms, hash tables, double-checked locking, RCU, hazard pointers), with discussion of performance and on the interaction between algorithm design and the underlying relaxed memory models. [3 blocks]
- **Research topics,** likely to include one hour on transactional memory and one guest lecture. [1 block]

Objectives

By the end of the course students should:

- have a good understanding of the semantics of concurrent programs, both at the multiprocessor level and the C/Java programming language level;
- have a good understanding of some key concurrent algorithms, with practical experience.
**Recommended reading**


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**Unit: Natural Language Processing**

This course is only taken by Part II 75% students.

*Lecturers: Professor S. Teufel and Dr P. Buttery*

*No. of lectures and practical classes: 12+3*

*Prerequisite courses: Machine Learning and Real-World Data, Formal Models of Language, Foundations of Data Science, Artificial Intelligence.*

*Capacity: 30*

**Aims**

This course introduces the fundamental techniques of natural language processing. It aims to explain the potential and the main limitations of these techniques. Some current research issues are introduced and some current and potential applications discussed and evaluated. Students will also be introduced to practical experimentation in natural language processing.

**Lectures**

- **Introduction.** Brief history of NLP research, some current applications, components of NLP systems.
- **Finite-state techniques.** Inflectional and derivational morphology, finite-state automata in NLP, finite-state transducers.
- **Prediction and part-of-speech tagging.** Corpora, simple N-grams, word prediction, stochastic tagging, evaluating system performance.
- **Context-free grammars and parsing.** Generative grammar, context-free grammars, parsing with context-free grammars, weights and probabilities. Some limitations of context-free grammars.
- **Dependency structures.** English as an outlier. Universal dependencies. Introduction to dependency parsing.
- **Compositional semantics.** Logical representations. Compositional semantics and lambda calculus. Inference and robust entailment. Negation.
- **Lexical semantics.** Semantic relations, WordNet, word senses.
• **Distributional semantics.** Representing lexical meaning with distributions. Similarity metrics.

• **Distributional semantics and deep learning.** Embeddings. Grounding. Multimodal systems and visual question answering.

• **Discourse processing.** Anaphora resolution, summarization.

• **Language generation and regeneration.** Generation and regeneration. Components of a generation system. Generation of referring expressions.

• **Recent NLP research.** Some recent NLP research.

• **Practical on sentiment analysis.** Students will build a sentiment analysis system which will be trained and evaluated on supplied data. The system will be built from existing components, but students will be expected to compare approaches and some programming will be required for this.

**Objectives**

By the end of the course students should:

- be able to discuss the current and likely future performance of several NLP applications;

- be able to describe briefly a fundamental technique for processing language for several subtasks, such as morphological processing, parsing, word sense disambiguation etc.;

- understand how these techniques draw on and relate to other areas of computer science.

**Recommended reading**

* Jurafsky, D. & Martin, J. (2008) *Speech and language processing*. Prentice Hall.
Principles of Communications

Lecturer: Professor J.A. Crowcroft
No. of lectures: 16
Suggested hours of supervisions: 4
Prerequisite course: Computer Networking
This course may be useful for the Part III course on Network Architectures.
Useful related courses: Computer Systems Modelling, Information Theory, Digital Signal Processing

Aims

This course aims to provide a detailed understanding of the underlying principles for how communications systems operate. Practical examples (from wired and wireless communications, the Internet, and other communications systems) are used to illustrate the principles.

Lectures

- **Introduction.** Course overview. Abstraction, layering. Review of structure of real networks, links, end systems and switching systems. [1 lecture]


- **Flow control and resource optimisation.** Control theory is a branch of engineering familiar to people building dynamic machines. It can be applied to network traffic. Stemming the flood, at source, sink, or in between? Optimisation as a model of network& user. TCP in the wild. [3 lectures]

- **Packet Scheduling.** Design choices for scheduling and queue management algorithms for packet forwarding, and fairness. [2 lectures]

- **The big picture for managing traffic.** Economics and policy are relevant to networks in many ways. Optimisation and game theory are both relevant topics discussed here. [2 lectures]

- **System Structures and Summary.** Abstraction, layering. The structure of real networks, links, end systems and switching. [2 lectures]

Objectives

At the end of the course students should be able to explain the underlying design and behaviour of protocols and networks, including capacity, topology, control and use. Several specific mathematical approaches are covered (control theory, optimisation).
Recommended reading


Background reading:

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Types

*Lecturer: Dr N. Krishnaswami*

*No. of lectures: 12*

*suggested hours of supervisions: 3*

*Prerequisite courses: Computation Theory, Semantics of Programming Languages*

**Aims**

The aim of this course is to show by example how type systems for programming languages can be defined and their properties developed, using techniques that were introduced in the Part IB course on *Semantics of Programming Languages*. The emphasis is on type systems for functional languages and their connection to constructive logic.

**Lectures**

- **Introduction.** The role of type systems in programming languages. Review of rule-based formalisation of type systems. [1 lecture]

- **Propositions as types.** The Curry-Howard correspondence between intuitionistic propositional calculus and simply-typed lambda calculus. Inductive types and iteration. Consistency and termination. [2 lectures]

- **Polymorphic lambda calculus (PLC).** PLC syntax and reduction semantics. Examples of datatypes definable in the polymorphic lambda calculus. Type inference. [3 lectures]

- **Monads and effects.** Explicit versus implicit effects. Using monadic types to control effects. References and polymorphism. Recursion and looping. [2 lectures]

- **Continuations and classical logic.** First-class continuations and control operators. Continuations as Curry-Howard for classical logic. Continuation-passing style. [2 lectures]

- **Dependent types.** Dependent function types. Indexed datatypes. Equality types and combining proofs with programming. [2 lectures]
Objectives

At the end of the course students should

- be able to use a rule-based specification of a type system to carry out type checking and type inference;
- understand by example the Curry-Howard correspondence between type systems and logics;
- understand how types can be used to control side-effects in programming;
- appreciate the expressive power of parametric polymorphism and dependent types.

Recommended reading

Lent Term 2020: Part II lectures

Unit: Cloud Computing

This course is only taken by Part II 75% students.

Lecturers: Dr E. Kalyvianaki and Dr A. Madhavapeddy

No. of lectures and practical classes: 10+3


Capacity: 50

Aims

This module aims to teach students the fundamentals of Cloud Computing covering topics such as virtualization, data centres, cloud resource management, cloud storage and popular cloud applications including batch and data stream processing. Emphasis is given on the different backend technologies to build and run efficient clouds and the way clouds are used by applications to realise computing on demand. The course will include practical tutorials on cloud infrastructure technologies. Students will be assessed via a Cloud-based coursework project.

Lectures

- Introduction to Cloud Computing
- Data centres
- Virtualization I
- Virtualization II
- MapReduce
- MapReduce advanced
- Resource management for virtualized data centres
- Cloud storage
- Cloud-based data stream processing

Objectives

By the end of the course students should:

- understand how modern clouds operate and provide computing on demand;
• understand about cloud availability, performance, scalability and cost;

• know about cloud infrastructure technologies including virtualization, data centres, resource management and storage;

• know how popular applications such as batch and data stream processing run efficiently on clouds;

Recommended reading

Marinescu, D.C. *Cloud Computing, Theory and Practice*. Morgan Kaufmann.

Comparative Architectures

*Lecturer: Dr R.D. Mullins*

*No. of lectures: 16*

*Suggested hours of supervisions: 4*

*Prerequisite course: Computer Design*

*Aims*

This course examines the techniques and underlying principles that are used to design high-performance computers and processors. Particular emphasis is placed on understanding the trade-offs involved when making design decisions at the architectural level. A range of processor architectures are explored and contrasted. In each case we examine their merits and limitations and how ultimately the ability to scale performance is restricted.
Lectures

- **Introduction.** The impact of technology scaling and market trends.
- **Fundamentals of Computer Design.** Amdahl’s law, energy/performance trade-offs, ISA design.
- **Advanced pipelining.** Pipeline hazards; exceptions; optimal pipeline depth; branch prediction; the branch target buffer [2 lectures]
- **Superscalar techniques.** Instruction-Level Parallelism (ILP); superscalar processor architecture [2 lectures]
- **Software approaches to exploiting ILP.** VLIW architectures; local and global instruction scheduling techniques; predicated instructions and support for speculative compiler optimisations.
- **Multithreaded processors.** Coarse-grained, fine-grained, simultaneous multithreading
- **The memory hierarchy.** Caches; programming for caches; prefetching [2 lectures]
- **Vector processors.** Vector machines; short vector/SIMD instruction set extensions; stream processing
- **Chip multiprocessors.** The communication model; memory consistency models; false sharing; multiprocessor memory hierarchies; cache coherence protocols; synchronization [2 lectures]
- **On-chip interconnection networks.** Bus-based interconnects; on-chip packet switched networks
- **Special-purpose architectures.** Converging approaches to computer design

Objectives

At the end of the course students should

- understand what determines processor design goals;
- appreciate what constrains the design process and how architectural trade-offs are made within these constraints;
- be able to describe the architecture and operation of pipelined and superscalar processors, including techniques such as branch prediction, register renaming and out-of-order execution;
- have an understanding of vector, multithreaded and multi-core processor architectures;
- for the architectures discussed, understand what ultimately limits their performance and application domain.
Recommended reading


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**Computer Vision**

*Lecturer: Professor J.G. Daugman*

*No. of lectures: 16*

*In lieu of supervisions, exercises will be set and reviewed in two Examples Classes.*

**Aims**

The aims of this course are to introduce the principles, models and applications of computer vision, as well as some mechanisms used in biological visual systems that may inspire design of artificial ones. The course will cover: image formation, structure, and coding; edge and feature detection; neural operators for image analysis; texture, colour, stereo, and motion; wavelet methods for visual coding and analysis; interpretation of surfaces, solids, and shapes; probabilistic classifiers; visual inference, recognition, and learning.

**Lectures**

- **Goals of computer vision; why they are so difficult.** Image formation, and the ill-posed problem of making 3D inferences about objects and their properties from images.

- **Image sensing, pixel arrays, CCD and CMOS cameras.** Image coding and information measures. Elementary operations on image arrays.

- **Biological visual mechanisms, from retina to cortex.** Photoreceptor sampling; receptive field profiles; stochastic impulse codes; channels and pathways. Neural image encoding operators.

- **Mathematical operations for extracting image structure.** Finite differences and directional derivatives. Filters; convolution; correlation. 2D Fourier domain theorems.

- **Edge detection operators; the information revealed by edges.** Gradient vector field; Laplacian operator and its zero-crossings.

- **Multi-scale contours, feature detection and matching.** SIFT (scale-invariant feature transform); pyramids. 2D wavelets as visual primitives. Active contours. Energy-minimising snakes.

- **Higher visual operations in brain cortical areas.** Multiple parallel mappings; streaming and divisions of labour; reciprocal feedback through the visual system.
• **Texture, colour, stereo, and motion descriptors.** Disambiguation and the achievement of invariances. Colour computation, motion and image segmentation.

• **Lambertian and specular surfaces; reflectance maps.** Geometric analysis of image formation from surfaces. Discounting the illuminant when inferring 3D structure and surface properties.

• **Shape representation.** Inferring 3D shape from shading; surface geometry. Boundary descriptors; codons. Object-centred volumetric coordinates.

• **Perceptual organisation and cognition.** Vision as model-building and graphics in the brain. Learning to see.

• **Lessons from neurological trauma and visual deficits.** Visual agnosias and illusions, and what they may imply about how vision works.

• **Bayesian inference in vision; knowledge-driven interpretations.** Classifiers, decision-making, and pattern recognition.

• **Model estimation.** Machine learning and statistical methods in vision.

• **Applications of machine learning in computer vision.** Discriminative and generative methods. Content based image retrieval.

• **Approaches to face detection, face recognition, and facial interpretation.** Cascaded detectors. Appearance *versus* model-based methods.

**Objectives**

At the end of the course students should

• understand visual processing from both “bottom-up” (data oriented) and “top-down” (goals oriented) perspectives;

• be able to decompose visual tasks into sequences of image analysis operations, representations, specific algorithms, and inference principles;

• understand the roles of image transformations and their invariances in pattern recognition and classification;

• be able to describe and contrast techniques for extracting and representing features, edges, shapes, and textures;

• be able to describe key aspects of how biological visual systems work; and be able to think of ways in which biological visual strategies might be implemented in machine vision, despite the enormous differences in hardware;

• be able to analyse the robustness, brittleness, generalizability, and performance of different approaches in computer vision;
• understand the roles of machine learning in computer vision today, including probabilistic inference, discriminative and generative methods;

• understand in depth at least one major practical application problem, such as face recognition, detection, or interpretation.

Recommended reading


Cryptography

Lecturer: Dr M.G. Kuhn
No. of lectures: 16
Suggested hours of supervisions: 3–4
Prerequisite courses: Mathematical Methods I from the NST Mathematics course, Discrete Mathematics, Complexity Theory

Aims

This course provides an overview of basic modern cryptographic techniques and covers essential concepts that users of cryptographic standards need to understand to achieve their intended security goals.

Lectures

• Cryptography. Overview, private vs. public-key ciphers, MACs vs. signatures, certificates, capabilities of adversary, Kerckhoffs’ principle.


• Private-key encryption. Stream ciphers, pseudo-random generators, attacking linear-congruential RNGs and LFSRs. Semantic security definitions, oracle queries, advantage, computational security, concrete-security proofs.

• Block ciphers. Pseudo-random functions and permutations. Birthday problem, random mappings. Feistel/Luby–Rackoff structure, DES, TDES, AES.

• Chosen-plaintext attack security. Security with multiple encryptions, randomized encryption. Modes of operation: ECB, CBC, OFB, CNT.

• Message authenticity. Malleability, MACs, existential unforgeability, CBC-MAC, ECBC-MAC, CMAC, birthday attacks, Carter-Wegman one-time MAC.
• **Authenticated encryption.** Chosen-ciphertext attack security, ciphertext integrity, encrypt-and-authenticate, authenticate-then-encrypt, encrypt-then-authenticate, padding oracle example, GCM.

• **Secure hash functions.** One-way functions, collision resistance, padding, Merkle–Damgård construction, sponge function, duplex construct, entropy pool, SHA standards.

• **Applications of secure hash functions.** HMAC, stream authentication, Merkle tree, commitment protocols, block chains, Bitcoin.

• **Key distribution problem.** Needham–Schroeder protocol, Kerberos, hardware-security modules, public-key encryption schemes, CPA and CCA security for asymmetric encryption.

• **Number theory, finite groups and fields.** Modular arithmetic, Euclid's algorithm, inversion, groups, rings, fields, GF($2^n$), subgroup order, cyclic groups, Euler's theorem, Chinese remainder theorem, modular roots, quadratic residues, modular exponentiation, easy and difficult problems. [2 lectures]

• **Discrete logarithm problem.** Baby-step-giant-step algorithm, computational and decision Diffie–Hellman problem, DH key exchange, ElGamal encryption, hybrid cryptography, Schnorr groups, elliptic-curve systems, key sizes. [2 lectures]

• **Trapdoor permutations.** Security definition, turning one into a public-key encryption scheme, RSA, attacks on “textbook” RSA, RSA as a trapdoor permutation, optimal asymmetric encryption padding, common factor attacks.

• **Digital signatures.** one-time signatures, ElGamal signatures, DSA, RSA signatures, Certificates, PKI.

**Objectives**

By the end of the course students should

• be familiar with commonly used standardized cryptographic building blocks;

• be able to match application requirements with concrete security definitions and identify their absence in naive schemes;

• understand various adversarial capabilities and basic attack algorithms and how they affect key sizes;

• understand and compare the finite groups most commonly used with discrete-logarithm schemes;

• understand the basic number theory underlying the most common public-key schemes, and some efficient implementation techniques.
Recommended reading


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**E-Commerce**

*Lecturers: Jack Lang, Stewart McTavish and others*

*No. of lectures: 8*

*Suggested hours of supervision: 2 (example classes if requested)*

*Prerequisite courses: Business Studies, Security, Economics, Law & Ethics*

**Aims**

This course aims to give students an outline of the issues involved in setting up an e-commerce site.

**Lectures**

- **The history of electronic commerce.** Mail order; EDI; web-based businesses, credit card processing, PKI, identity and other hot topics.

- **Network economics.** Real and virtual networks, supply-side versus demand-side scale economies, Metcalfe’s law, the dominant firm model, the differentiated pricing model Data Protection Act, Distance Selling regulations, business models.

- **Web site design.** Stock and price control; domain names, common mistakes, dynamic pages, transition diagrams, content management systems, multiple targets.

- **Web site implementation.** Merchant systems, system design and sizing, enterprise integration, payment mechanisms, CRM and help desks. Personalisation and internationalisation.

- **The law and electronic commerce.** Contract and tort; copyright; binding actions; liabilities and remedies. Legislation: RIP; Data Protection; EU Directives on Distance Selling and Electronic Signatures.

- **Putting it into practice.** Search engine interaction, driving and analysing traffic; dynamic pricing models. Integration with traditional media. Logs and audit, data mining modelling the user. Collaborative filtering and affinity marketing brand value, building communities, typical behaviour.

- **Finance.** How business plans are put together. Funding Internet ventures; the recent hysteria; maximising shareholder value. Future trends.

**Objectives**

At the end of the course students should know how to apply their computer science skills to the conduct of e-commerce with some understanding of the legal, security, commercial, economic, marketing and infrastructure issues involved.

**Recommended reading**


Additional reading:


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**Machine Learning and Bayesian Inference**

*Lecturer: Dr S.B. Holden*

*No. of lectures:* 12

*Suggested hours of supervisions:* 3

*Prerequisite courses:* Artificial Intelligence, Foundations of Data Science, Discrete Mathematics and Probability, Linear Algebra and Calculus from the NST Mathematics course.

**Aims**

The Part 1B course *Artificial Intelligence* introduced simple neural networks for supervised learning, and logic-based methods for knowledge representation and reasoning. This course has two aims. First, to provide a rigorous introduction to machine learning, moving beyond the supervised case and ultimately presenting state-of-the-art methods. Second, to provide an introduction to the wider area of probabilistic methods for representing and reasoning with knowledge.

**Lectures**

• **How to classify optimally.** Treating learning probabilistically. Bayesian decision theory and Bayes optimal classification. Likelihood functions and priors. Bayes theorem as applied to supervised learning. The maximum likelihood and maximum *a posteriori* hypotheses. What does this teach us about the backpropagation algorithm? [2 lectures]

• **Linear classifiers I.** Supervised learning via error minimization. Iterative reweighted least squares. The maximum margin classifier. [1 lecture]

• **Support vector machines (SVMs).** The kernel trick. Problem formulation. Constrained optimization and the dual problem. SVM algorithm. [2 lectures]


• **Linear classifiers II.** The Bayesian approach to neural networks. [1 lecture]

• **Unsupervised learning I.** The *k*-means algorithm. Clustering as a maximum likelihood problem. [1 lecture]

• **Unsupervised learning II.** The EM algorithm and its application to clustering. [1 lecture]

• **Bayesian networks I.** Representing uncertain knowledge using Bayesian networks. Conditional independence. Exact inference in Bayesian networks. [1 lecture]

• **Bayesian networks II.** Markov random fields. Approximate inference. Markov chain Monte Carlo methods. [1 lecture]

**Objectives**

At the end of this course students should:

• Understand how learning and inference can be captured within a probabilistic framework, and know how probability theory can be applied in practice as a means of handling uncertainty in AI systems.

• Understand several algorithms for machine learning and apply those methods in practice with proper regard for good experimental practice.

**Recommended reading**

If you are going to buy a single book for this course we recommend:


The course text for Artificial Intelligence I:


covers some relevant material but often in insufficient detail. Similarly:
gives a gentle introduction to some of the course material, but only an introduction.
Recently a few new books have appeared that cover a lot of relevant ground well. For example:


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**Unit: Mobile Robot Systems**

This course is only taken by Part II 75% students.

*Lecturers: Dr A. Prorok*

*No. of lectures and practical classes: 16*

*Prerequisite courses: NST Mathematics, Artificial Intelligence, Algorithms.*

*Capacity: 40*

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**Aims**

This course teaches the foundations of autonomous mobile robots, covering topics such as perception, motion control, and planning. It also teaches algorithmic strategies that enable the coordination of multi-robot systems and robot swarms. The course will feature several practical sessions with hands-on robot programming. The students will undertake mini-projects, which will be formally evaluated through a report and presentation.

**Lectures**

- **Robot motion and control.** Kinematics, control models, trajectory tracking.
- **Control architectures.** Sensor-actuator loops, reactive path planning.
- **Sensing.** Sensors, perception.
- **Localization.** Markov localization, environment modeling, SLAM.
- **Navigation.** Planning, receding horizon control.
- **Multi-robot systems I.** Centralization vs. decentralization, robot swarms.
- **Multi-robot systems II.** Consensus algorithms, graph-theoretic methods.
- **Multi-robot systems III.** Task assignment.
- **Multi-robot systems IV.** Multi-robot path planning.
Objectives

By the end of the course students should:

- understand how to control a mobile robot;
- understand how a robot perceives its environment;
- understand how a robot plans actions (navigation paths);
- know paradigms of coordination in systems of multiple robots;
- know classical multi-robot problems and their solution methods;

Recommended reading


Mobile and Sensor Systems

Lecturer: Prof C. Mascolo, Dr R.K. Harle
No. of lectures: 12
Suggested hours of supervisions: 3
Prerequisite courses: Operating Systems, Concurrent and Distributed Systems

Aims

This course will cover topics in the areas of mobile systems and communications, and sensor systems and sensor networking. It aims to help students develop and understand the additional complexity introduced by mobility and sensing, including energy constraints, communication in dynamic networks and handling measurement errors. The course will be using various applications to exemplify concepts.

Lectures

- Introduction to Mobile Systems. MAC Layer concepts. Examples of mobile systems, differences with non mobile systems. Introduction to MAC layer protocols of wireless and mobile systems.
• **Mobile Infrastructure Communication and Opportunistic Networking.**
  Description of common communication architectures and protocols for mobile and introduction to models of opportunistic networking.

• **Introduction to Sensor Systems, MAC Layer concepts and Internet of Things.**
  Sensor systems challenges and applications. Concepts related to duty cycling and energy preservation protocols. Emerging concepts and communications protocols for the Internet of Things.

• **Sensor Systems Routing Protocols.** Communication protocols, data aggregation and dissemination in sensor networks.


• **Mobile Sensing: Systems Considerations** Considerations of energy preservation. Local computation vs cloud computation. Storage and Latency.

• **Privacy in Mobile and Sensor Systems.** Concepts of mobile and sensor systems privacy. Privacy and sensor based activity inference. Mobility prediction and privacy.

• **Localization I.** Basic concepts. Proximity, trilateration, triangulation, ToA, TDoA. Examples including ultrasonic, mobile networks, UWB.

• **Localization II.** GNSS, RSS fingerprinting/WiFi positioning.

• **Tracking.** Kalman filtering, particle filtering, Pedestrian Dead-Reckoning as an example.

• **Mobile/Wearable health sensing.** Health as a key driver for wearables. High availability and low fidelity (wearable) vs low availability and high fidelity (clinical). Sensing challenges (low power, noisy, poor contact). PPG case study. False positives and the base rate fallacy.

• **Robots and Drones** Concepts related to control, communication and coordination of robotic systems.

**Objectives**

On completing the course, students should be able to

• describe similarities and differences between standard distributed systems and mobile and sensor systems;

• explain the fundamental tradeoffs related to energy limitations and communication needs in these systems;

• argue for and against different mobile and sensor systems architectures and protocols.

• Understand typical error sources for sensing and be aware of techniques to minimise them.
• put concepts into context of current applications of mobile and sensor systems as described in the course.

Recommended reading

The course is based mainly on research papers cited in each lecture. The following books, however, contain some of the more traditional concepts.


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**Optimising Compilers**

*Lecturer: Dr T.M. Jones*

*No. of lectures: 16*

*Suggested hours of supervisions: 4*

*Prerequisite course: Compiler Construction*

**Aims**

The aims of this course are to introduce the principles of program optimisation and related issues in decompilation. The course will cover optimisations of programs at the abstract syntax, flowgraph and target-code level. It will also examine how related techniques can be used in the process of decompilation.

**Lectures**

- **Introduction and motivation.** Outline of an optimising compiler. Optimisation partitioned: analysis shows a property holds which enables a transformation. The flow graph; representation of programming concepts including argument and result passing. The phase-order problem.


- **Classical dataflow analysis.** Graph algorithms, live and avail sets. Register allocation by register colouring. Common sub-expression elimination. Spilling to memory; treatment of CSE-introduced temporaries. Data flow anomalies. Static Single Assignment (SSA) form.

- **Higher-level optimisations.** Abstract interpretation, Strictness analysis. Constraint-based analysis, Control flow analysis for lambda-calculus. Rule-based inference of program properties, Types and effect systems. Points-to and alias analysis.
• **Target-dependent optimisations.** Instruction selection. Instruction scheduling and its phase-order problem.

• ** Decompilation.** Legal/ethical issues. Some basic ideas, control flow and type reconstruction.

**Objectives**

At the end of the course students should

• be able to explain program analyses as dataflow equations on a flowgraph;

• know various techniques for high-level optimisation of programs at the abstract syntax level;

• understand how code may be re-scheduled to improve execution speed;

• know the basic ideas of decompilation.

**Recommended reading**


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**Unit: Probability and Computation**

This course is only taken by Part II 75% students.

*Lecturers: Dr. T.M. Sauerwald, Dr. N. Rivera, Dr. J. Sylvester, Dr. L. Zanetti*

*No. of lectures and practical classes: 16*

*Prerequisite courses: Algorithms, Foundations of Data Science.*

*Capacity: 30*

**Aims**

The aim of this course is to introduce the design and analysis of randomised algorithms. It starts by introducing some essential tools and ideas from probability and graph theory, and develops this knowledge through analysing a variety of examples of randomised algorithms and processes. Ultimately the course demonstrates that randomness can be an elegant programming technique, and particularly helpful when time or space are restricted.
Lectures


- **Concentration Inequalities**: How to derive Chernoff bounds, Applications to load balancing and quick-sort. Extensions of Chernoff bounds.

- **Random walks and Markov chains**: Transition matrices, Stationary distribution and Convergence. Mixing time and total variation distance. Applications of Markov chains such as Connectivity testing, Solving 2-SAT, Sampling from unknown distributions and MCMC.

- **Spectral Analysis of Random Walks**: Convergence Rate, Cheeger’s Inequality, Applications including Graph Clustering.


Objectives

By the end of the course students should be able to:

- learn how to use randomness in the design of algorithms;

- apply randomisation to various problems coming from optimisation, machine learning and distributed computing;

- use results from probability theory to analyse the performance of randomised algorithms.

Recommended reading


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Quantum Computing

Lecturer: Dr. S. Herbert

No. of lectures: 16

Suggested hours of supervisions: 4

Prerequisite courses: Foundations of Data Science, Computation Theory
Aims

The principal aim of the course is to introduce students to the basics of the quantum model of computation. The model will be used to study algorithms for searching, factorisation and quantum chemistry as well as other important topics in quantum information such as cryptography and super-dense coding. Issues in the complexity of computation will also be explored. A second aim of the course is to introduce student to near-term quantum computing. To this end, error-correction and adiabatic quantum computing are studied.

Lectures

- **Bits and qubits.** Introduction to quantum states and measurements with motivating examples. Comparison with discrete classical states.

- **Linear algebra.** Review of linear algebra: vector spaces, linear operators, Dirac notation, the tensor product.

- **The postulates of quantum mechanics.** Postulates of quantum mechanics, incl. evolution and measurement.

- **Important concepts in quantum mechanics.** Entanglement, distinguishing orthogonal and non-orthogonal quantum states, no-cloning and no signalling.

- **The quantum circuit model.** The circuit model of quantum computation. Quantum gates and circuits. Universality of the quantum circuit model, and efficient simulation of arbitrary two-qubit gates with a standard universal set of gates.

- **Some applications of quantum information.** Applications of quantum information (other than quantum computation): quantum key distribution, superdense coding and quantum teleportation.

- **Deutsch–Jozsa algorithm.** Introducing Deutsch’s problem and Deutsch’s algorithm leading onto its generalisation, the Deutsch–Jozsa algorithm.

- **Quantum search.** Grover’s search algorithm: analysis and lower bounds.

- **Quantum Fourier Transform & Quantum Phase Estimation.** Definition of the Quantum Fourier Transform (QFT), and efficient representation thereof as a quantum circuit. Application of the QFT to enable Quantum Phase Estimation (QPE).

- **Application 1 of QFT / QPE: Factoring.** Shor’s algorithm: reduction of factoring to period finding and then using the QFT for period finding.

- **Application 2 of QFT / QPE: Quantum Chemistry.** Efficient simulation of quantum systems, and applications to real-world problems in quantum chemistry.

- **Quantum complexity.** Quantum complexity classes and their relationship to classical complexity. Comparison with probabilistic computation.

- **Quantum error correction.** Introducing the concept of quantum error correction required for the following lecture on fault-tolerance.
- **Fault tolerant quantum computing.** Elements of fault tolerant computing; the threshold theorem for efficient suppression of errors.

- **Adiabatic quantum computing.** The quantum adiabatic theorem, and adiabatic optimisation. Quantum annealing and D-Wave.

- **Case studies in near-term quantum computation.** Examples of state-of-the-art quantum algorithms and computers, including superconducting and networked quantum computers.

**Objectives**

At the end of the course students should:

- understand the quantum model of computation and the basic principles of quantum mechanics;
- be familiar with basic quantum algorithms and their analysis;
- be familiar with basic quantum protocols such as teleportation and superdense coding;
- see how the quantum model relates to classical models of deterministic and probabilistic computation.
- appreciate the importance of efficient error-suppression if quantum computation is to yield an advantage over classical computation.
- gain a general understanding of the important topics in near-term quantum computing, including adiabatic quantum computing.

**Recommended reading**

**Books:**


**Papers:**


Aharonov D., Quantum computation [arXiv:quant-ph/9812037]

McCardle S. et al, Quantum computational chemistry [https://arxiv.org/abs/1808.10402]

Other lecture notes:
Umesh Vazirani (UC Berkeley): [http://www-inst.eecs.berkeley.edu/~cs191/sp12/]
John Preskill (Caltech): [http://www.theory.caltech.edu/people/preskill/ph229/]
Andrew Childs (University of Maryland): [http://cs.umd.edu/~amchilds/qa/]
John Watrous (University of Waterloo): [https://cs.uwaterloo.ca/~watrous/TQI/]

Unit: Topics in Concurrency

This course is only taken by Part II 75% students.

Lecturer: Professor G. Winskel
No. of lectures: 12
Suggested hours of supervisions: 3

Prerequisite course: Semantics of Programming Languages (specifically, an idea of operational semantics and how to reason from it)

Capacity: no restrictions

Aims

The aim of this course is to introduce fundamental concepts and techniques in the theory of concurrent processes. It will provide languages, models, logics and methods to formalise and reason about concurrent systems.

Lectures

- **Simple parallelism and nondeterminism.** Dijkstra’s guarded commands. Communication by shared variables: A language of parallel commands. [1 lecture]


- **Introduction to Petri nets.** Petri nets, basic definitions and concepts. Petri-net semantics of CCS. [1 lecture]

• **Mobile computation.** An introduction to process languages with process passing and name generation. [2 lectures]

**Objectives**

At the end of the course students should

• know the basic theory of concurrent processes: non-deterministic and parallel commands, the process language CCS, its transition-system semantics, bisimulation, the modal mu-calculus, Petri nets, languages for cryptographic protocols and mobile computation;

• be able to formalise and to some extent analyse concurrent processes: establish bisimulation or its absence in simple cases, express and establish simple properties of transition systems in the modal mu-calculus, argue with respect to a process language semantics for secrecy or authentication properties of a small cryptographic protocol, formalise mobile computation.

**Recommended reading**

Comprehensive notes will be provided.

Further reading:


Easter Term 2020: Part II lectures

Advanced Algorithms

Lecturer: Dr T.M. Sauerwald

No. of lectures: 12

Suggested hours of supervisions: 3

Prerequisite courses: Algorithms

Aims

The aim of this course is to introduce advanced techniques for the design and analysis of algorithms that arise in a variety of applications. A particular focus will be on parallel algorithms, linear programming and approximation algorithms.

Lectures

- **Sorting Networks.** Zero-one principle. Merging Network, Bitonic Sorter. Counting Networks. [CLRS2, Chapter 27]

- **Linear Programming.** Definitions and Applications. Formulating Linear Programs. The Simplex Algorithm. Finding Initial Solutions. [CLRS3, Chapter 29]

- **Approximation Algorithms.** (Fully) Polynomial-Time Approximation Schemes. Design Techniques. Applications: Vertex Cover, Subset-Sum, Parallel Machine Scheduling, Travelling Salesman Problem (including a practical demonstration how to solve a TSP instance exactly using linear programming), Hardness of Approximation. [CLRS3, Chapter 35]


Objectives

At the end of the course students should

- have an understanding of algorithm design for parallel computers;

- be able to formulate, analyse and solve linear programs;

- have learned a variety of tools to design efficient (approximation) algorithms.
Recommended reading


Business Studies Seminars

*Lecturer: Jack Lang, Stewart McTavish and others*

No. of seminars: 8

Aims

This course is a series of seminars by former members and friends of the Laboratory about their real-world experiences of starting and running high technology companies. It is a follow on to the Business Studies course in the Michaelmas Term. It provides practical examples and case studies, and the opportunity to network with and learn from actual entrepreneurs.

Lectures

Eight lectures by eight different entrepreneurs.

Objectives

At the end of the course students should have a better knowledge of the pleasures and pitfalls of starting a high tech company.

Recommended reading


See also the additional reading list on the Business Studies web page.