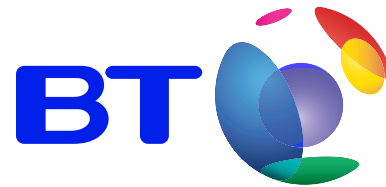


MSc projects at BT Research

Keith Briggs Keith.Briggs@bt.com

`keithbriggs.info`



University of Bath MSc talk 2007 Dec 12 1400

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Adastral Park, Martlesham, Suffolk



- ▶ Cambridge-Ipswich high-tech corridor
- ▶ 2000 technologists
- ▶ 15 companies
- ▶ UCL, Univ of Essex

BT Research centres

- ▶ Broadband Centre
- ▶ Foresight Centre
- ▶ IT futures research Centre
- ▶ Intelligent systems Centre
- ▶ Mobility Centre
- ▶ Networks Centre
- ▶ Pervasive ICT Centre
- ▶ Security Centre
- ▶ Asian Research Centre
- ▶ Disruptive lab at MIT

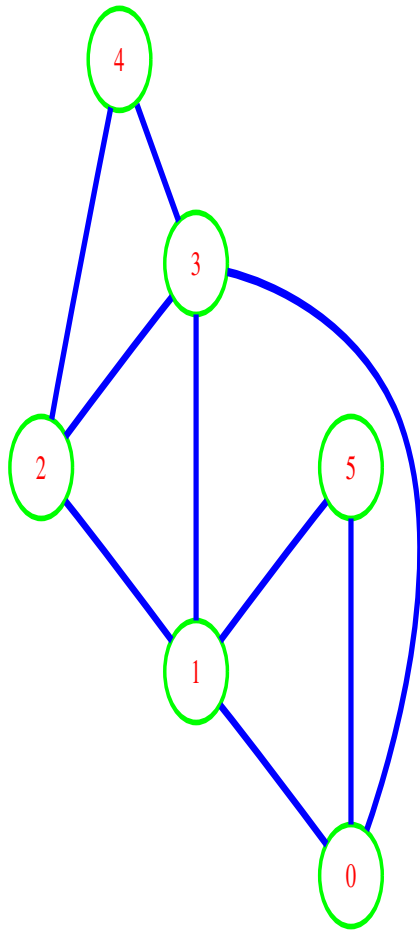
MSc projects supervised by Keith Briggs

Details: www.keithbriggs.info/MSc_project_ideas_2008.html

Mostly in discrete maths, two in ODEs. . .

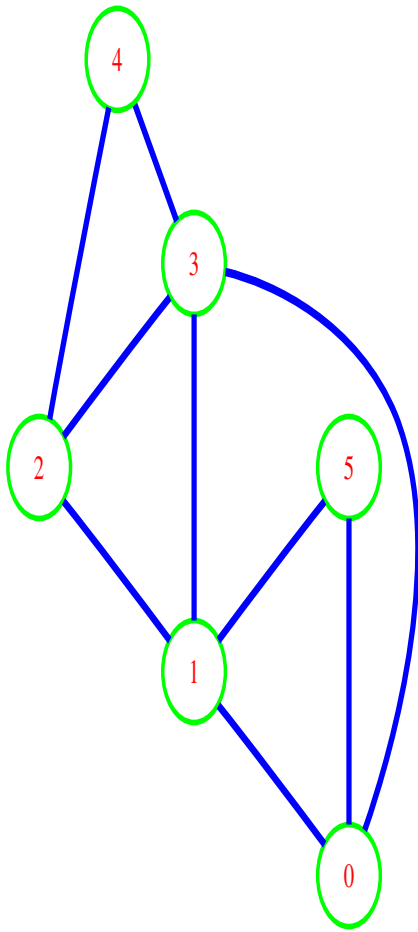
- ▷ 1. *Semidefinite programming and graph theory*
- ▷ 2. *Singular value decomposition updating*
- ▷ 3. *Singular value decomposition methods for Lyapunov exponents*
- ▷ 4. *Automatic differentiation (AD) methods for ODE sensitivity analysis*
- ▷ 5. *Fast random selection*
- ▷ 6. *Random sampling of set partitions*
- ▷ 7. *Fast counting*
- ▷ 8. *Random sampling of unlabelled structures*
- ▷ 9. *Convex optimization in python*
- ▷ 10. *Geographical computations*
- ▷ 11. *Statistics of Roman roads in Britain*

Graph concepts

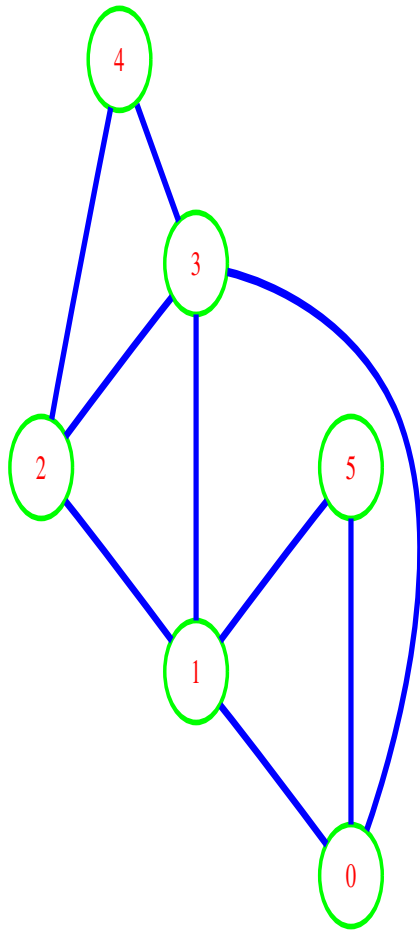


Graph concepts

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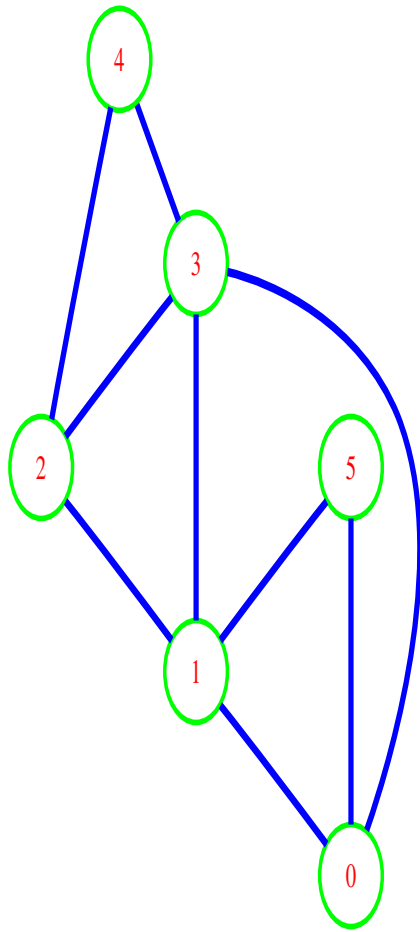


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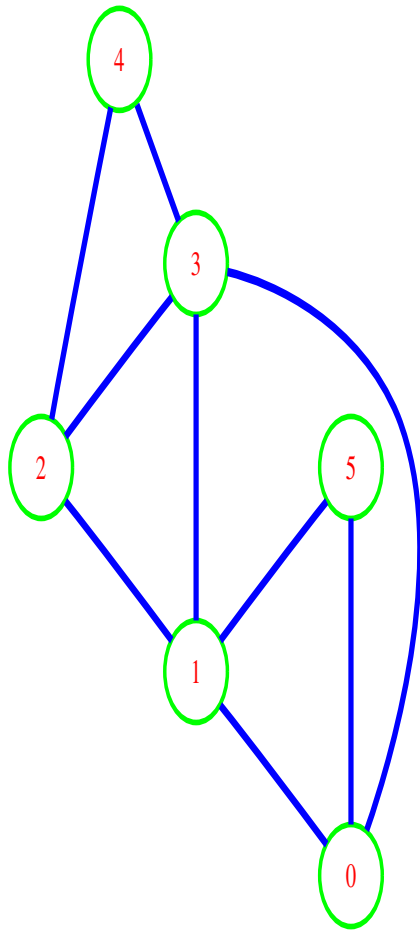
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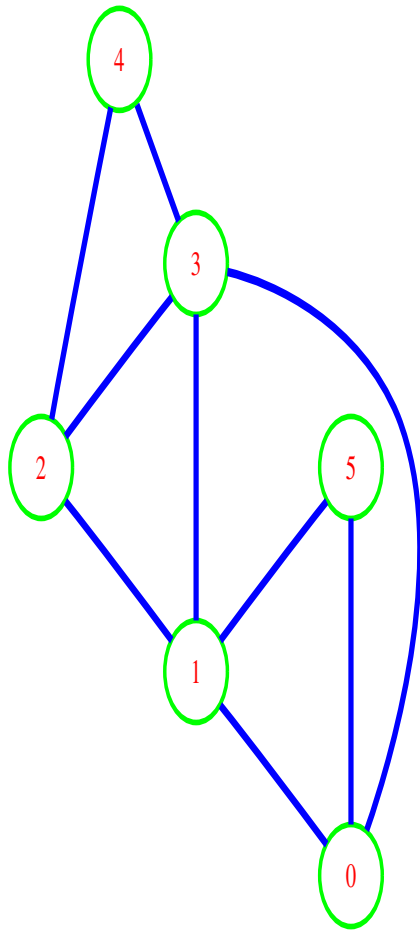
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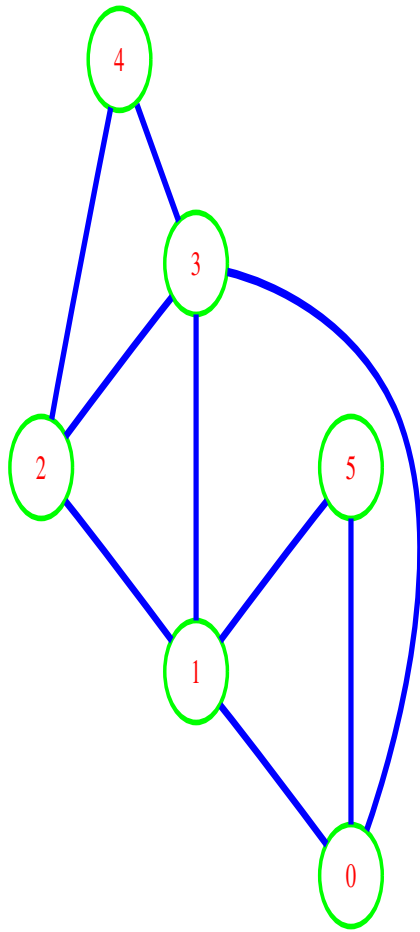
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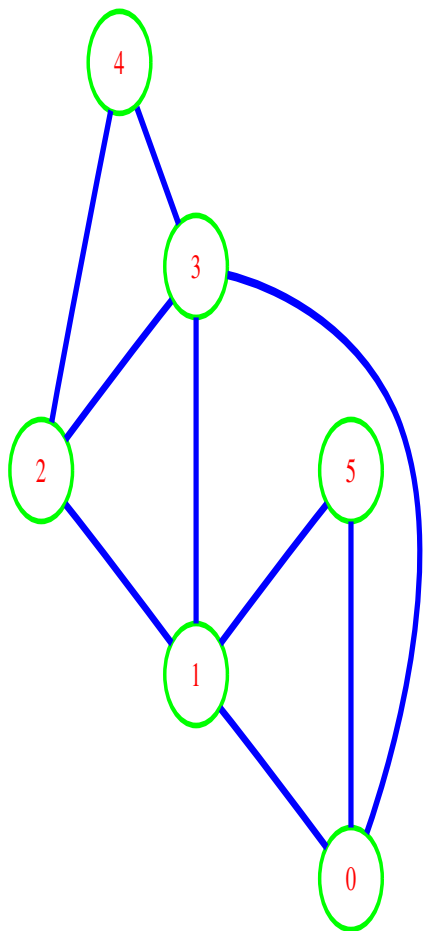
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Semidefinite programming and graph theory

Semidefinite programming (SDP) is a kind of generalized linear programming.

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- ▶ Such problems include the Lovasz theta number (a lower bound for the chromatic number), the maximal stable set problem, and the maximum cut problem.
- ▶ An important outcome of the project would be a determination of how the computation time scales with problem size. Geometric optimization problems could also be studied.

2. Singular value decomposition updating

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- ▶ However, I have an application in signal processing, where only the approximate SVD is required, but it is required to rapidly update the factors after a small change in the matrix (perturbation theory, if you like).
- ▶ I have an idea how to do this, and the project would be to develop the idea and implement it in software.

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- ▶ I have much experience in AD and can supply software for this step.

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- ▶ C++ programming is needed (as operator overloading is necessary).

5. Fast random selection

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- ▶ This project would investigate these methods and check their efficiency in practice. There are applications to the generation of large random graphs.

6. Random sampling of set partitions

- ▶ A partition of $[n] = \{1, 2, \dots, n\}$ is an assignment each element to a subset called a block, without regard to the labelling of the block, or order of elements within a block. Thus for $[5]$, $01|2|34$ and $2|10|43$ are the same partition.

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- ▶ Beyond about $n = 15$, there are too many partitions to allow us to construct all of them in reasonable time.
- ▶ Thus, for large sets, simulations must rely on a uniform random sampling procedure. This project would investigate such methods and produce a well-designed C library for use in other projects.

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- ▶ The algorithm should have practical applications in informatics; for example, counting the number of different packet types in a network.

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- ▶ The resulting algorithms generate in an unbiased manner discrete configurations that may have nontrivial symmetries, and they do so by means of real-arithmetic computations.
- ▶ This project would develop software for some of these methods, and measure its efficiency in practice. There are important applications in statistical computing.

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- ▶ This project would survey the field of convex optimization and investigate and evaluate this software. We would hope to try applications in the field of radio technology.

10. Geographical computations

- ▶ In this project, I developed an algorithm for computing intervisibility of points on the earth's surface, using elevation data from a shuttle flight.

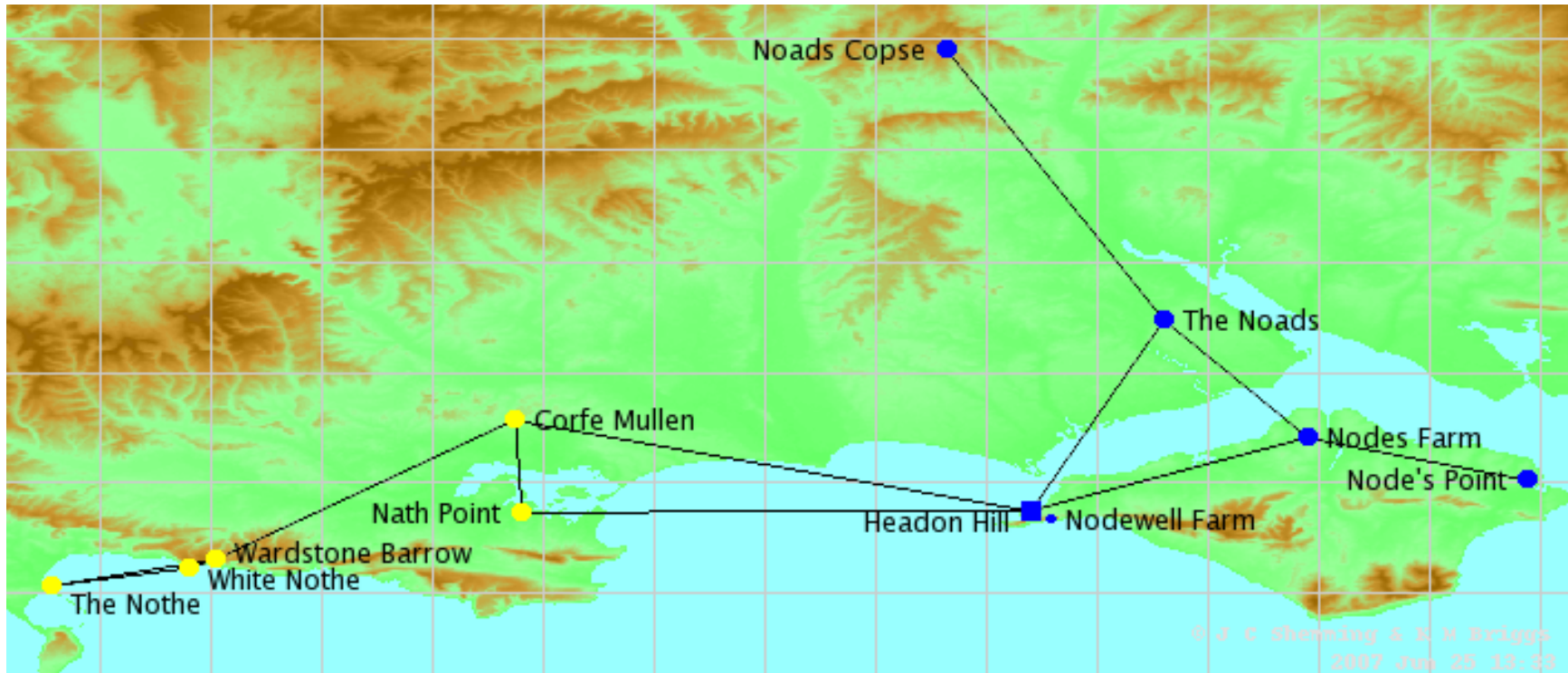
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- ▶ The application is to the historical study of ancient defence systems.

10. Geographical computations ctnd.



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- ▶ The maths involved is interesting computational geometry, and fast algorithms are needed for things like the point-in-polygon test, and for distance to the nearest of a given finite set of straight line segments.

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