

Matlab

Linear Algebra etc.

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Please Interrupt

This course assumes a fair amount of background

1: that you already happy using **Matlab**

E.g. **basic use**, **syntax** and **error handling**

2: that you already know some **linear algebra**

At least up to elementary use of **matrices**

It will refer to a bit more, but will explain

- If you don't understand, **please interrupt**

Don't feel afraid to ask any question you want to

Beyond the Course

Email scientific-computing@ucs for advice

- Ask almost anything about scientific computing

<http://www-uxsup.csx.cam.ac.uk/courses/...>
[.../Matlab/](http://www-uxsup.csx.cam.ac.uk/courses/.../Matlab/)

<http://www.mathworks.com/access/helpdesk/...>
[.../help/techdoc/](http://www.mathworks.com/access/helpdesk/.../help/techdoc/)

What Is Linear Algebra?

Could call it the arithmetic of **matrices**
It's more general than you might think

Need to explain some mathematics
Don't Panic – it will be over-simplified!

You can do a great deal in Matlab

- As always, follow the motto “**festina lente**”
“**Make haste slowly**” – i.e. start with simple uses

Structure Of Course

- Overview of what analyses are possible
- Basic **matrix facilities** in Matlab
- **Real** and **complex** linear algebra
- Summary of more **advanced** matrix facilities

What Are Matrices?

Effectively a **2-D** rectangular grid of **elements**

1.2	2.3	3.4	4.5
5.6	6.7	7.8	8.9
9.0	0.1	1.2	2.3

1-D arrays are also called **vectors**

n-D arrays may be called **tensors**

Won't cover them, because poor in Matlab

Yes, mathematicians, I know – over-simplification!

Terminology

Some people use **matrix** to mean only **2-D**

Others talk about **matrices** of any **rank**

The standard term is **matrix algebra** not **array algebra**

Matlab tends to use the term **array** generically

- But it also uses **array** for other purposes
 - This course often uses **matrix** generically
- But uses **vector** or **n-D** where it matters

Elements of Matrices

- These are not limited to **real numbers**
Can actually belong to any mathematical **field**
Examples:

- **Real** (\mathbb{R}) or **complex** (\mathbb{C}) numbers
- Ratios of **integers** (**rational numbers**)
- Ratios of **polynomials/multinomials**
- And more

Course covers **real** but mentions **complex**
You use them in almost exactly the same way

Reminder

123456789 is an integer

12345.6789 is a real number

123.45+678.9i is a complex number

`complex(123.45,678.9)` creates the same number

What Can We Do?

All of basic **matrix arithmetic**, obviously
Including some quite complicated operations

Solution of **simultaneous linear equations**

Eigenvalues and **eigenvectors**

Matrix **decompositions** of quite a few types

Plus (with more hassle) their **error analysis**

Fourier transforms are just **linear algebra**, too

Physics, Chemistry etc.

Anything expressible in normal **matrix notation**

- That's almost everything, really!

But that isn't always **practically** possible

Matlab is slower than **NAG**, but you can call **NAG**

- Working with **expressions** is **not** covered

You need to use **Mathematica** for that

But you can often get much more information

Statistical Uses

- Regression and analysis of variance
- Multivariate probability functions

Calculating the **errors** is the tricky bit
It's **NOT** the same as in most physics!

- Also **Markov processes** – finite state machines
This is where transitions are **probabilistic**
Working with these is just more matrix algebra
- Standard textbooks give the matrix formulae
You just carry on from there ...

Matlab and Matrices

Will describe how Matlab provides them

And explain how to construct and display them

And perform other basic matrix operations

Matrix Notation (1)

Conventional layout of a **4x3** matrix **A**
Multiplied by a **3** vector

$$\begin{array}{ccc} 11 & 12 & 13 \\ 21 & 22 & 23 \\ 31 & 32 & 33 \\ 41 & 42 & 43 \end{array} \quad \times \quad \begin{array}{c} 7 \\ 8 \\ 9 \end{array} \quad \rightarrow \quad \begin{array}{c} 290 \\ 530 \\ 770 \\ 1010 \end{array}$$

$A_{3,2}$ is the value **32**

530 is **$21 \times 7 + 22 \times 8 + 23 \times 9$**

Matrix Notation (2)

Now we do the same thing in Matlab

All of the details will be explained in due course

```
a = [ 11 , 12 , 13 ; 21 , 22 , 23 ; 31 , 32 , 33 ; 41 , 42 , 43 ]
```

```
    11    12    13
```

```
    21    22    23
```

```
    31    32    33
```

```
    41    42    43
```

```
b = [ 7 ; 8 ; 9 ]
```

```
a ( 3 , 2 ) -> 32
```

```
a * b -> [ 290 , 530 , 770 , 1010 ]
```

Notation in Papers

There are a zillion – one for each sub–area
'Standard' tensor notation has changed, too
Here is another over–simplification

A_i , A^i , \bar{A} or \tilde{A} is a vector

A_i may also refer to element i of vector A

B_{ij} or B_i^j is a matrix

$A_i B_{ij}$ often means $\sum_i A_i \times B_{ij}$

Algorithms may use $A(i)$ or $A[i]$ and $B(i,j)$ or $B[i,j]$

Matrices as Lists (1)

We can input a **list of values**

```
a = [ 9.8 , 8.7 , 7.6 , 6.5 , 5.4 , 4.3 , 3.2 , 2.1 , 1.0 ]
```

```
b = [ 1.2 , 2.3 , 3.4 , 4.5 ; 5.6 , 6.7 , 7.8 , 8.9 ;  
      9.0 , 0.1 , 1.2 , 2.3 ]
```

```
1.2  2.3  3.4  4.5  
5.6  6.7  7.8  8.9  
9.   0.1  1.2  2.3
```

You use **commas** (,) between numbers in **rows**
And **semicolons** (;) between successive **columns**

Matrices as Lists (2)

Values need not be simple numbers

Expressions using defined variables are allowed

$$x = 1.23$$

$$y = 4.56$$

$$c = [1 + x^2, x * y ; -x * y, 1 + y^2]$$

$$\begin{array}{cc} 2.5129 & 5.6088 \\ -5.6088 & 21.7936 \end{array}$$

Row Major or Column Major?

I find those terms **seriously** confusing

We want to know which subscript varies fastest

- Matlab like **Mathematica** and **C**, **NOT Fortran**

First subscript (**rows**) varies **slowest**

Last subscript (**columns**) varies **fastest**

```
a = [ 11 , 12 , 13 ; 21 , 22 , 23 ; 31 , 32 , 33 ; 41 , 42 , 43 ]
```

```
a ( 3 , 2 )  ->  32
```

11 , 12 , 13 is the first of **4 rows**

11 , 21 , 31 , 41 is the first of **3 columns**

Vectors

A **vector** can be either a **row** or **column**

- And you need to use the right one

$$a = [1.2 \ , \ 2.3 \ , \ 3.4]$$
$$1.2 \quad 2.3 \quad 3.4$$

$$b = [1.2 \ ; \ 2.3 \ ; \ 3.4]$$
$$1.2$$
$$2.3$$
$$3.4$$

$$b'$$
$$1.2 \quad 2.3 \quad 3.4$$

Index Ranges (1)

a : b means all values from **a** to **b**

```
a = [ 1 , 2 , 3 , 4 , 5 , 6 , 7 , 8 , 9 , 10 ]
```

```
a ( 5 : 7 ) -> [ 5 , 6 , 7 ]
```

```
a = [ 1 , 2 , 3 ; 4 , 5 , 6 ; 7 , 8 , 9 ]
```

```
a ( 1 : 2 , 2 : 3 ) -> [ 2 , 3 ; 5 , 6 ]
```

Index Ranges (2)

Just a **:** means all values in a **row** or **column**
Can be combined with **indices** and **index ranges**

```
a = [ 1 , 2 , 3 ; 4 , 5 , 6 ; 7 , 8 , 9 ]
```

```
a ( : , 2 : 3 )
```

```
2 3  
5 6  
8 9
```

```
a ( 2 , : ) -> [ 4 , 5 , 6 ]
```

```
4 5 6
```

Matrix Constructors (1)

`zeros (2)` \rightarrow `[0 , 0 ; 0 , 0]`

`zeros (1 , 2)` \rightarrow `[0 , 0]`

`zeros (2 , 1)` \rightarrow `[0 ; 0]`

`1.23 * ones (2)` \rightarrow `[1.23 , 1.23 ; 1.23 , 1.23]`

`rand (2)` \rightarrow `[0.9649 , 0.9706 ; 0.1576 , 0.9572]`

`eye (3)` \rightarrow `[1 , 0 , 0 ; 0 , 1 , 0 ; 0 , 0 , 1]`

Most have several options and extra features

Matrix Constructors (2)

There are also many specialised standard matrices
These are Hilbert and inverse Hilbert ones

hilb (3)

1.0000	0.5000	0.3333
0.5000	0.3333	0.2500
0.3333	0.2500	0.2000

invhilb (3)

9	-36	30
-36	192	-180
30	-180	180

Also companion, Hankel, Toeplitz matrices, etc.

- Look them up when and if you need them!

Matrix Constructors (3)

You can construct **linear sequence** vectors

```
linspace ( - 2.3 , 8.9 , 7 )  
-2.3000  -0.4333  1.4333  3.3000  5.1667  
          7.0333  8.9000
```

And then assign them to **rows** or **columns**
Try this and see what it does:

```
a = zeros ( 7 , 3 ) ;  
a ( : , 1 ) = linspace ( 1 , 7 , 7 ) ;  
a ( : , 2 ) = linspace ( - 2.3 , 8.9 , 7 ) ;  
a ( : , 3 ) = linspace ( 5.6 , 7.8 , 7 ) ;
```

Advanced Construction

Matlab has **lots** of more advanced features
If you can't do what you want, look for something

```
a = [ 11 , 12 , 13 ; 21 , 22 , 23 ; 31 , 32 , 33 ]  
[ a , a , a ]  
[ a ; a ; a ]  
repmat ( a , 5 , 7 )
```

Look in **Elementary Matrices and Arrays**
And in **Array Manipulation**

Importing Data

Unfortunately, Matlab isn't brilliant at this
This course won't go into how to do it
General advice:

- Use **load** and **save** for storage
Uses Matlab's own format – version dependent
- Use **importdata** for numbers in text form
Input them in the format that Matlab expects
- Use **Python** (or **Perl**) for other formats
Write the data out in **importdata** format

Basic Functions (1)

`disp(A)` displays `A` – compare that with `A`

```
a = [ 11 , 12 , 13 ]
```

```
a
```

```
    a =
```

```
    11    12    13
```

```
disp(a)
```

```
    11    12    13
```

Basic Functions (2)

ndims is the **rank** (number of dimensions)
size gives the actual **dimensions**

```
a = [ 11 , 12 , 13 ;   21 , 22 , 23 ]
```

```
ndims ( a )  
      2
```

```
size ( a )  
      2  3
```

A Vector 'Gotcha'

The **rank** of a **vector** is **one** (1), right?

- Not usually in Matlab, it isn't
Most '**vectors**' are actually **degenerate matrices**

```
a = [ 11 , 12 , 13 ]  
b = [ 11 ; 12 ; 13 ]
```

```
ndims ( a )  ->  2  
size ( a )   ->  1  3  
ndims ( b )  ->  2  
size ( b )   ->  3  1
```

IEEE 754 Features

Matlab has a large collection of **IEEE 754** functions
Matlab's support is **not IEEE 754**-compliant
Nor is **Java**'s, or most other languages'

- I **strongly** recommend **NOT** using them
See the course **How Computers Handle Numbers**

The reasons are far too complicated to go into here
But not even **experts** can use them **reliably**
Please ask if you want to know more details

Basic Operations

$A + B$ addition

$A - B$ subtraction

$A * B$ **matrix** multiplication

$A ^ K$ **K**th power of **matrix**

$A '$ normal transpose (**conjugated**)

$A .'$ transpose (**non-conjugated**)

All of those work as in normal **mathematics**
Generally, use only **positive integer** power **K**

Matrix Multiplication

Remember that this is **NOT** commutative
 $A*B$ is generally not equal to $B*A$

Try the following if you need to convince yourself

```
A = rand(5)
```

```
B = rand(5)
```

```
A*B
```

```
B*A
```

And the **product** of two **symmetric** matrices
will usually be **unsymmetric**

Matrix Division

A / B matrix (right) division

$A .\setminus B$ matrix left division

- Take care about numerical problems with these
Serious ones with both accuracy and range
- They are effectively linear equation solvers
We will discuss that in more detail later

It is easy to get confused when using these

Elementwise Operations

You can also do **elementwise** operations

$A .* B$ **elementwise** multiplication

$A .^K$ **elementwise** K th power

$A ./ B$ **elementwise** (right) division

$A .\ B$ **elementwise** left division

You can use any power that makes sense

Diagonalisation (1)

Can construct a **diagonal** matrix from a **vector**
Or convert a **matrix** into its **diagonal** vector

`diag ([1.23 , 4.56]) -> [1.23 , 0 ; 0 , 4.56]`

`diag ([1.23 ; 4.56]) -> [1.23 , 0 ; 0 , 4.56]`

`b = [1.2 , 2.3 , 3.4 ; 4.5 , 5.6 , 6.7 ; 7.8 , 8.9 , 9.0]`

`diag (b) -> [1.2 ; 5.6 ; 9.0]`

That **dual use** is counter-intuitive, but this is useful:

`diag (diag (b))`

Diagonalisation (2)

You can also construct **block diagonal** matrices
The **block matrices** can be of any shape

```
a = [ 11 , 12 , 13 ;    21 , 22 , 23 ;    31 , 32 , 33 ]
```

```
b = [ 100 , 200 ;    300 , 400 ]
```

```
c = 5
```

```
blkdiag ( a , b , c , c , c , b , a )
```

I **strongly** recommend experimenting before use
Especially if any of your **blocks** aren't **square**

More Functions (1)

max and **min** collapse along one **dimension**

```
a = [ 11 , 12 , 13 ; 21 , 22 , 23 ; 31 , 32 , 33 ]
```

```
max ( a )  
    31    32    33
```

```
min ( a )  
    11    12    13
```

```
max ( a , [ ] , 2 )  
    13  
    23  
    33
```

More Functions (2)

`prod` and `sum` use a different `syntax`

```
a = [ 11 , 12 , 13 ;   21 , 22 , 23 ;   31 , 32 , 33 ]
```

```
sum ( a )
```

```
    63    66    69
```

```
prod ( a )
```

```
    7161    8448    9867
```

```
sum ( a , 2 )
```

```
    36
```

```
    66
```

```
    96
```

More Functions (3)

Most **basic mathematical** functions work **elementwise**

```
a = [ 1 , 2 , 3 ]  
b = [ -2 , 0 , 1 ]
```

```
sqrt ( a )
```

```
1.0000  1.4142  1.7321
```

```
exp ( a )
```

```
2.7183  7.3891  20.0855
```

```
atan2 ( a , b )
```

```
2.6779  1.5708  1.2490
```

arrayfun can be used for **your** functions

More Functions (4)

The usual **inner** and **cross** products of **vectors**
`dot(V1,V2)` and `cross(V1,V2)`

But no **outer** product nor **normalisation**

You have to program those yourself, by hand; e.g.

```
a = [ 1 , 2 , 3 ]
```

```
b = a / sqrt ( dot ( a , a ) )
```

```
0.2673  0.5345  0.8018
```

Other absences for **multi-dimensional** matrices

Practical Break

We shall now stop for some **practical exercises**
These are mainly for inexperienced Matlab users

- I don't care **which** system you use
Just don't expect any help with **Microsoft!**

I use **matlab -nodisplay** under **Linux**

Numeric Linear Algebra

This is defined only for **real** and **complex**
Matlab uses **IEEE 754 64-bit** – **c.15** sig. figs

- This has some special mathematics to itself
Can do a **lot** more than for general matrices
- Are going to cover only the simplest analyses
There's a **huge** amount more in Matlab
And **even** more in books and papers

Determinant

`det` gives the **determinant**

As usual, watch out for **numerical problems**

```
det ( hilb ( 10 ) ) -> 2.1644e-53
```

```
det ( invhilb ( 10 ) ) -> 4.6207e+52
```

So far, so good. Now let's try a harder problem

```
det ( hilb ( 20 ) ) -> -1.1004e-195
```

```
det ( invhilb ( 20 ) ) -> 3.1788e+254
```

Norms

`norm` gives a choice of induced norms

Calculate other norms using the usual formulae

`norm(A)` is the 2-norm of A

I.e. the largest singular value of A

`norm(A,1)` is the 1-norm or largest column sum

`norm(A,inf)` is the ∞ -norm or largest row sum

`norm(A,'fro')` is the Frobenius norm

`normest(A)` is an approximation to `norm`

`norm(V,p)` is the p -norm for vector V and integer p

Condition Numbers

$\text{cond}(A)$ is the 2-norm condition number

$\text{cond}(A, \text{opt})$ is the opt-norm condition number

opt can be anything allowed in norm

$\text{rcond}(A)$ is LAPACK approx. reciprocal cond. no

$\text{condeig}(A)$ is the vector of eigenvalue cond. nos

Relevant only for unsymmetric matrices

Other Functions

`trace(A)` is the sum of diagonal elements

`expm(A)` is the **matrix exponential**

`inv(A)` is the **matrix inverse**

You **very** rarely should be using this

rank, **null**, **orth** etc. for subspaces

Most of these are for relatively advanced use

- Watch out for **serious** numerical **'gotchas'**

Linear Equations (1)

In the common, simple cases, **Just Do It**

$$a = \begin{bmatrix} 4.2 & 2.2 & -3.9 & 9.3 & 0.1 \\ 8.6 & 0.0 & 0.7 & -2.3 & -0.3 \\ 8.4 & -5.9 & -8.1 & 9.6 & 3.8 \\ -0.8 & -9.4 & -9.9 & 9.9 & 5.0 \\ -1.3 & -8.1 & 0.6 & -9.2 & -7.3 \end{bmatrix}$$

$$b = [-6.8, 2.3, 2.7, -7.0, 2.0]$$

`linsolve (a , b')`

$$[1.45411, -12.4949, 24.5078, 11.8408, 0.422917]$$

Linear Equations (2)

How do these relate to **matrix division**?

Yes, I find this very confusing, too!

```
a = rand ( 5 )
```

```
b = rand ( 5 )
```

```
linsolve ( a , b )
```

```
a \ b
```

```
( b' / a' )'
```

All of the above are more-or-less equivalent

Linear Equations (3)

How did I work that out? Trying every combination ...
I then worked out what Matlab's authors were thinking

- Always run a **cross-check** when doing that
I.e. check you have done the **right** calculation

$$a * \text{linsolve}(a, b) \rightarrow b$$

$$(b / a) * a \rightarrow b$$

Linear Equations (4)

It's **critical** to look for its **warnings**

```
linsolve ( [ 1.2 , 3.4 ; 2.4 , 6.8 ] , [ 1.0 ; 1.0 ] )
```

Warning: Matrix is singular to working precision.

-Inf

Inf

```
linsolve ( hilb ( 20 ) , rand( 20 , 1 ) )
```

Warning: Matrix is close to singular or badly scaled.

Results may be inaccurate. RCOND = 1.995254e-19.

Linear Equations (5)

- If any, the results are often complete nonsense
They may be NaNs, infinities, zeroes
⇒ Or plausible values that are just plain wrong

- Matlab won't stop by itself on a warning
You have to do the checking, by hand

The command `dbstop if warning` looks useful here
But works only if you are running the debugger!

- Naturally, this applies to all warnings
Not just ones that occur in linear equations

Decompositions

You can get just the decompositions

The main ones are $\text{chol}(A)$ and $\text{lu}(A)$

Cholesky for real positive definite symmetric only

LU for pretty well any square matrix

$\text{qr}(A)$ is also used for eigenvalues

$\text{pinv}(A)$ is the Moore–Penrose pseudo–inverse

Least Squares Solutions (1)

Matlab uses these for **over-determined** problems
Generally, everything does what you expect

```
a = rand ( 10 , 3 )  
b = rand ( 10 , 1 )  
linsolve ( a , b )
```

```
0.5731  
0.1851  
0.2837
```

You will get different **random results**, of course

Least Squares Solutions (2)

`lsq` can handle known **covariances**

Look it up if you need to do that

But what if you need to do more **advanced** work?

Including any serious **statistics** or **regression**

- Use the Matlab **statistics toolbox** (extra cash)
- Statistical package, like **Genstat** or **SPSS**
- Program the **formulae** yourself, if you can

Fourier Transforms (1)

Matlab doesn't call them **linear algebra**

Look under **Data Analysis** for **fft***

```
a = [ -0.92 , 9.1 , 2.3 , 5.7 , 4.9 , -2.8 , -5.6 ,  
      6.7 , -7.0 , 9.0 ]
```

```
b = fftn ( a )
```

```
21.4 , 11.8 - 14.1i , -4.6 + 3.8i , 9.9 - 5.2i ,  
-15.4 + 15.9i , -34.0 , -15.4 - 15.9i ,  
9.9 + 5.2i , -4.6 - 3.8i , 11.8 + 14.1i
```

```
ifftn ( b )
```

```
-0.92 , 9.1 , 2.3 , 5.7 , 4.9 , -2.8 , -5.6 ,  
6.7 , -7.0 , 9.0
```


Fourier Transforms (2)

- `fftn` and `ifftn` work on **matrices**, too
They are the **multi-dimensional** array forms

`fft/ifft` and `fft2/ifft2` look more obvious

- But they have a very serious ‘**gotcha**’
Unexpected behaviour on **too many** dimensions

`fft/ifft` work on the **first** dimension only

I.e. they apply multiple **1-dimensional** FFTs

I haven't experimented with what `fft2/ifft2` do

Fourier Transforms (3)

You can make `fft/ifft` work on **another** dimension

Using the same nasty syntax as for `max/min`

The following two operations are **equivalent** – try them

```
a = rand ( 5 )
```

```
fft2 ( a )
```

```
fft ( fft ( a , [ ] , 2 ) )
```

Eigenanalysis (1)

- Things start to get a bit hairier, here
That is because the **mathematics** does
- All **square** matrices have all **eigenvalues**
But **real** matrices may have **complex** eigenvalues
- All **real symmetric** matrices have all **eigenvectors**
And those are always **orthogonal**
The same applies to all **complex Hermitian** ones
- But **unsymmetric** matrices may be nasty

Eigenanalysis (2)

Simple use is, er, simple

```
a = [ 4.2 , 2.2 , -3.9 , 9.3 , 0.1 ;  
      8.6 , 0.0 , 0.7 , -2.3 , -0.3 ;  
      8.4 , -5.9 , -8.1 , 9.6 , 3.8 ;  
      -0.8 , -9.4 , -9.9 , 9.9 , 5.0 ;  
      -1.3 , -8.1 , 0.6 , -9.2 , -7.3 ]
```

```
eig ( a )
```

```
6.4585 + 9.8975i  
6.4585 - 9.8975i  
-7.2840 + 4.4546i  
-7.2840 - 4.4546i  
0.3510
```

Characteristic Polynomial

Eigenvalues are the **roots** of that

You can calculate it directly, if you want

```
a = [ 4.2 , 2.2 , -3.9 , 9.3 , 0.1 ;  
      8.6 , 0.0 , 0.7 , -2.3 , -0.3 ;  
      8.4 , -5.9 , -8.1 , 9.6 , 3.8 ;  
      -0.8 , -9.4 , -9.9 , 9.9 , 5.0 ;  
      -1.3 , -8.1 , 0.6 , -9.2 , -7.3 ]
```

```
poly ( a )
```

```
1.0e+03 *
```

```
0.0010 0.0013 0.0238 1.0845 9.7983 -3.5741
```

Eigenvectors (1)

You can get the **eigenvectors** almost as easily

If you don't know the following **syntax**, just use it
It just assigns **two** results, rather than **one**

```
a = [ 4.2 , 2.2 , -3.9 ; 8.6 , 0.3 , 0.7 ;  
      8.4 , -5.9 , -8.1 ]
```

```
[ p , q ] = eig ( a )
```

Returns the **eigenvectors** in the **columns** of **p**
And the **eigenvalues** as a **diagonal matrix** in **q**

Eigenvectors (2)

The output is omitted because of verbosity – try it

Let's check that we have got the right answer

$p * q / p$ should be the matrix a

```
      a = ...
4.2000    2.2000   -3.9000
8.6000    0.3000    0.7000
8.4000   -5.9000   -8.1000
      p * q / p
4.2000 + 0.0000i    2.2000 - 0.0000i   -3.9000
8.6000 + 0.0000i    0.3000 - 0.0000i    0.7000 - 0.0000i
8.4000 + 0.0000i   -5.9000 - 0.0000i   -8.1000
```

Eigenvectors (3)

Here, **really** don't rely on the diagnostics

$$a = \begin{bmatrix} 1.0 & 1.0 \\ 0.0 & 1.0 \end{bmatrix}$$

$$[p, q] = \text{eig}(a)$$

$$p * q / p$$
$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Oops! Not even a **warning**

Eigenvectors (4)

Let's try another of the same level of **nastiness**

$$a = [0.0 , 1.0 ; 0.0 , 0.0]$$

$$[p , q] = \text{eig} (a)$$

$$p * q / p$$

Warning: Matrix is close to singular or badly scaled.
Results may be inaccurate. RCOND = 2.004168e-292.

$$\begin{array}{cc} 0 & 0 \\ 0 & 0 \end{array}$$

More Eigenanalysis (1)

There are a lot of extra **features** and **functions** I haven't investigated, and it's not one of my areas

Matlab says **eig** has problems with **sparse** matrices
There's another function, **eigs**, which **may** do better

You can get the **Hessenberg** and **Schur** forms

And then there's the question of whether to **balance**

More Eigenanalysis (2)

- Real symmetric and complex Hermitian are easy
You will be unlucky to have problems with those ones
Mere numerical inaccuracy excepted, of course
- Otherwise: Be Very Careful
Unsymmetric eigenanalysis is full of 'gotchas'
- Many experts say you shouldn't do it at all
You should use Singular value decomposition

Singular Values (1)

Singular value decomposition is called SVD
A more robust extension of eigenanalysis

Gives the same results in the simple cases
The term for these is normal matrices

Also handles non-square matrices
And ones with missing eigenvectors

If you don't know it, don't worry about it
But it's an important technique in many fields

Singular Values (2)

Try the following with a variety of matrices **a**

`eig (a)`
`svd (a)`

`[p , q] = eig (a)`
`p * q / p -> a`

`[r , s , t] = svd (a)`
`r * s * t' -> a`

Singular Values (3)

The trivial cases give exactly the same answers
Matlab uses the reverse orders of **eigenvalues**

```
a = hilb ( 3 )
```

```
eig (a)
```

```
[ 0.0027 ; 0.1223 ; 1.4083 ]
```

```
svd ( a )
```

```
[ 1.4083 ; 0.1223 ; 0.0027 ]
```

Singular Values (4)

But **SVD** handles the foul cases I used **correctly**

```
a = [ 1.0 , 1.0 ; 0.0 , 1.0 ]
```

```
[ r , s , t ] = svd ( a )
```

```
r * s * t'
```

```
1.0000 1.0000
```

```
0.0000 1.0000
```

```
a = [ 0.0 , 1.0 ; 0.0 , 0.0 ]
```

```
[ r , s , t ] = svd ( a )
```

```
r * s * t'
```

```
0 1
```

```
0 0
```

Complex Numbers (1)

FFT and eigenanalysis have produced these
We have completely ignored them – why?

- Because there has been nothing to say!
Really do use complex numbers just like real ones
Their differences don't show up in linear algebra

⇒ There is only one critical warning
Regard complex exceptions as pure poison
That means mainly overflow and division by zero

See the course [How Computers Handle Numbers](#)

Complex Numbers (2)

$$\begin{aligned} a &= [4.2 + 2.2i , -3.9 + 9.3i, 0.1 + 0.0i ; \\ & \quad 8.6 + 0.0i , 0.7 - 2.3i , 0.0 - 0.3i ; \\ & \quad 8.4 - 5.9i , -8.1 + 9.6i , 3.8 - 0.8i] \\ b &= [-6.8 + 2.3i ; 2.7 - 7.0i ; 2.0 + 0.0i] \end{aligned}$$

$$\begin{aligned} \text{linsolve} (a , b) \\ & 0.0362 - 0.5311i \\ & 0.7195 + 0.6147i \\ & 4.0269 + 1.5706i \end{aligned}$$

$$\begin{aligned} \text{eig} (a) \\ & -3.2517 - 8.1715i \\ & 7.9506 + 7.7844i \\ & 4.0011 - 0.5129i \end{aligned}$$

A Bit of Numerical Analysis

Very roughly, the error in linear algebra is:

$$N \times \text{cond. number} \times \text{epsilon}$$

Where N is the size of the matrix

Cond. number is how 'nasty' the matrix is

epsilon is the error in the values/arithmetic

- Almost always, the main error is in the **input data**

Good linear algebra **algorithms** are very accurate

⇒ Rounding error isn't usually the problem

Real vs Floating-Point

See “How Computers Handle Numbers”

Only significant problem is loss of accuracy

Not going to teach much numerical analysis

But it's well-understood for much of linear algebra

- PLEASE don't use single precision

And do watch out for growth of inaccuracy

Solution of Equations (1)

Let's look at a classic numerically foul problem
The **Hilbert matrix** is **positive definite**
And horribly **ill-conditioned** ...

In **Mathematica rational** arithmetic, result is exact

```
a = HilbertMatrix [ 10 ]
```

```
b = ConstantArray [ 1 , 10 ]
```

```
c = LinearSolve [ a , b ]
```

```
{ -10 , 990 , -23760 , 240240 , -1261260 , 3783780 ,  
  -6726720 , 7001280 , -3938220 , 923780 }
```

Solution of Equations (2)

{ -10 , 990 , -23760 , 240240 , -1261260 , 3783780 ,
-6726720 , 7001280 , -3938220 , 923780 }

Now we do it in **Matlab's floating-point**

```
a = hilb ( 10 )  
b = ones (10 , 1 )  
num2str ( linsolve ( a , b ) , 6 )
```

-9.99803 , 989.83 , -23756.4 , 240207 ,
-1.2611e+06 , 3.78335e+06 , -6.72601e+06 ,
7.0006e+06 , -3.93786e+06 , 923701

Error Analysis

Traditionally, this is overall error analysis

Usually in terms of **norms** etc.

It is a well-understood area, with useful results

- Use the formulae for it in books etc.

Sorry, but there is no time to go into it further

Sparse Arrays (1)

Matlab has good support for **sparse** arrays
You use them just like **dense** arrays – ha, ha!

- **Don't be fooled** – this is a **problem area**
It can be done very well, but don't just rush in

First problem is many operations cause **infilling**
A **16 MB** array can expand to **16 GB**

- **Sparse algorithms** are designed to minimise that

Sparse Arrays (2)

Problem is that nothing comes for free

- Very often the sacrifice is **numerical robustness**
Some problems may give **inaccurate** answers, or **fail**
I mentioned Matlab's warning about the **eig** function

So proceed **cautiously** with sparse arrays

- Find out what **best practice** is before proceeding

But **lots** of people do it, very **successfully**

Esoteric Analyses

The above are the **basic methods** of linear algebra

- Every field has its own **specialised methods**
There are **thousands** of them and **each field differs**
- Matlab has a **few**, otherwise **program** them
Almost always **formulae** in **reference** books
- The only warning is to watch out for **errors**
Both **algorithm failure** and **numerical inaccuracy**

Practical Break

We shall now stop for some **practical exercises**
These use the actual **linear algebra** functions

- The same comments apply as before

But, before that, one more slide ...

Feedback

Please fill in your **green forms**

I am particularly interested in the following:

- Would you have liked an **intermediate** course?
- Would you like a course on **sparse arrays**?
- Would you like a course on the **NAG** toolbox?
- Any other related courses that you would like?
On Matlab or any other **scientific computing**