

Computational Foundations of Cognitive Science

Lecture 14: Inverses and Eigenvectors in Matlab; Plotting and Graphics

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Inverse

The command `inv(A)` computes the inverse of A . Matlab complains if the matrix is singular:

```
> A = [1 2 3; 2 5 3; 1 0 8]; B = [1 6 4; 2 4 -1; -1 2 5];
> disp(inv(A));
   -40    16     9
    13    -5    -3
     5     -2    -1
> disp(inv(B));
warning: inverse: matrix singular to machine precision
```

We can test the property $AA^{-1} = I$:

```
> disp(inv(A) * A);
   1     0     0
   0     1     0
   0     0     1
```

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Reading: McMahon, Ch. 3

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Inverse

We can test a few more properties of the inverse, such as $(AB)^{-1} = B^{-1}A^{-1}$ and $(A^T)^{-1} = (A^{-1})^T$:

```
> format rat;
> B = [1 6 4; 2 4 -1; -1 2 5];
> disp(inv(A * B));
   88/3    -209/18    -119/18
  -34/3    163/36     91/36
   -1/3     1/9
> disp(inv(B) * inv(A));
   88/3    -209/18    -119/18
  -34/3    163/36     91/36
   -1/3     1/9     1/9
> disp(inv(A'))';
  -40     13     5
   16     -5     -2
    9     -3     -1
```

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Determinant

The command `det(A)` computes the determinant of A :

```
> A = [1 2 3; 2 5 3; 1 0 8]; B = [1 6 4; 2 4 -1; -1 2 5];
> disp(det(A));
-1
> disp(det(B));
0
```

Recall that $\det(B) = 0$ indicates that B is singular (not invertible). To compute the inverse based on the determinant:

```
> A = [1 2; 2 5]; disp(inv(A));
5 -2
-2 1
> Ai = 1/det(A) * [A(2, 2) -A(2, 1); -A(1, 2) A(1, 1)];
> disp(Ai);
5 -2
-2 1
```



Eigenvalues

Use `eig(A)` to obtain the eigenvalues of A :

```
> A = [1 3; 4 2];
> disp(eig(A));
-2
5
```

Let's check this against the characteristic equation of A :

```
> disp(-2 * eye(2) - A);
-3 -3
-4 -4
> disp(det(-2 * eye(2) - A));
0
```

Recall that the determinant of the characteristic equation of A has to be zero.



Eigenvectors

`[X, L] = eig(A)` returns a matrix X that contains the eigenvectors, and a matrix L that contains the eigenvalues of A :

```
> [X, L] = eig(A);
> disp(X);
-0.7071 -0.6000
0.7071 -0.8000
> disp(L);
-2 0
0 5
```

Note that Matlab scales the eigenvectors so that the norm of each vector is one. To avoid that, use the `nobalance` option:

```
> [X, L] = eig(A, 'nobalance'); disp(X);
-1.0000 -0.7500
1.0000 -1.0000
```



Eigenvectors

Let's check if these vectors are really eigenvectors. They have to have the property $Ax = \lambda x$:

```
> disp(A * X(:,1));
2
-2
> disp(L(1,1) * X(:,1));
2
-2
```

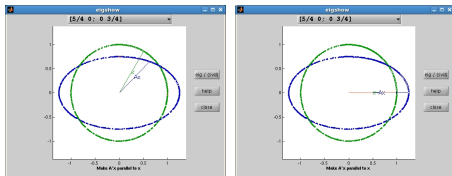
Note that the eigenvectors for A actually involve a scaling factor:

$\begin{bmatrix} -t \\ t \end{bmatrix}$ and $\begin{bmatrix} \frac{3}{4}t \\ t \end{bmatrix}$, but Matlab instantiates t .



Eigenvectors

Matlab's eigshow is a good way of getting an intuition for how eigenvectors work:



Plotting Functions

The `plot(x, y)` command in Matlab plots two vectors **x** and **y** against each other, with **x** representing the values on the x-axis and **y** representing the corresponding values on the y-axis.

The x-values can be generated with `x = [start:interval:end]`, which generates a vector with values ranging from `start` to `end`, spaced using `interval`.

We can then apply a function to **x** and call `plot`:

```
> x = [0:0.1:10];
> disp(x);
0.00 0.10 0.20 0.30 ... 10.00
> y = cos(x);
> disp(y);
1.00 0.99 0.98 0.95 ... -0.83
> plot(x, y);
```

Mid-lecture Problem

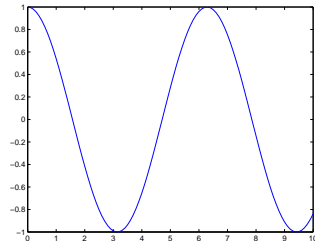
For a matrix **A**, assume that **X** is a matrix that contains the eigenvectors of **A**, and **Λ** is a matrix containing the eigenvalues of **A** on the diagonal.

Use Matlab to show that:

$$A = X\Lambda X^{-1}$$

What is this decomposition useful for?

Plotting Functions



Plotting Functions

The command `plot(x, y)` plots the content of arbitrary vectors. Functions can also be plotted using `fplot(function_string, [start end])`. This automatically chooses an optimal interval.

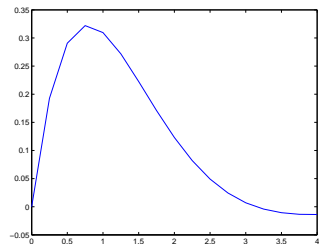
Compare:

```
> x = [0:0.25:4];
> y = exp(-x) .* sin(x);
> plot(x, y);
```

with:

```
> fplot('exp(-x) * sin(x)', [0, 4]);
```

Plotting Functions



Plotting Options

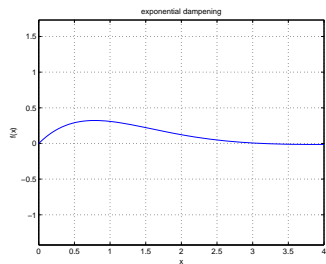
A range of options can be appended to `plot` or `fplot`:

```
xlabel label of x-axis
ylabel label of y-axis
title graph title string
legend graph legend string
grid switch grid on or off
axis axis spacing can be square or equal
    or [xmin xmax ymin ymax]
```

Example:

```
> fplot('exp(-x) * sin(x)', [0, 4], xlabel('x'),
    ylabel('f(x)'), title('exponential dampening'),
    grid on, axis equal;
```

Plotting Functions



Plotting Options

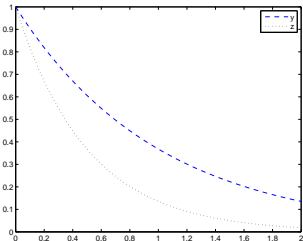
We can plot more than one function in the same graph, simply by giving plot multiple arguments:

```
> x = [0 : 0.01 : 5];
> y = exp(-x);
> z = exp(-2*x);
> plot(x, y, '--', x, z, ':'), legend('y', 'z'),
axis([0 2 0 1]);
```

Here, the third argument specifies the line type: '-' for straight line, '--' for dashed line, ':' for dotted line.

Note also the use of the legend option to introduce a legend, and the axis option to specify axis spacing.

Plotting Functions



Plotting Discrete Data

The plot command can be used to plot discrete data as well, but Matlab also offers a number of special graph types for this.

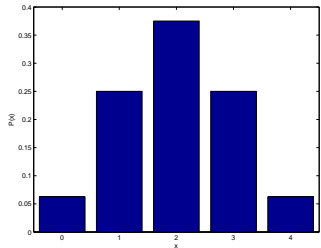
Assume we have the following probability distribution (probability $P(x)$ of obtaining x head when tossing a coin four times):

x	0	1	2	3	5
$P(x)$	$\frac{1}{16}$	$\frac{4}{16}$	$\frac{6}{16}$	$\frac{4}{16}$	$\frac{1}{16}$

Plot this distribution as a bar chart:

```
> x = [0 : 4];
> y = [1/16 4/16 6/16 4/16 1/16];
> bar(x, y), xlabel('x'), ylabel('P(x)');
```

Plotting Functions



Plotting Discrete Data

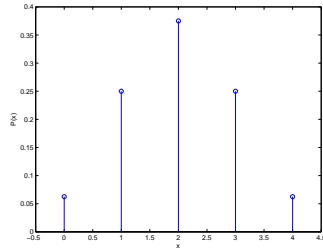
Plot the same data as a stem plot or as a scatter plot:

```
> stem(x, y, xlabel('x'), ylabel('P(x)'),
axis([-0.5 4.5 0 0.4]));
> plot(x, y, 'o'), xlabel('x'), ylabel('P(x)'),
axis([-0.5 4.5 0 0.4]);
```

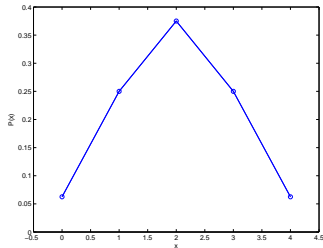
We can superimpose multiple graphs by saying hold. For example, we can use this connect the dots in the scatter plot:

```
> plot(x, y, 'o'), xlabel('x'), ylabel('P(x)'),
> hold;
> plot(x, y);
```

Plotting Functions



Plotting Functions



Processing Images

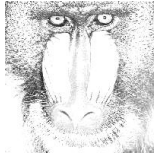
In Matlab, images can be processed as matrices. For example, a grayscale image of size 200×200 pixels is a matrix of integers ranging from 0 (black) to 255 (white).

Images can be read from a file using `imread`, saved to a file using `imwrite`, and displayed using `imshow`.

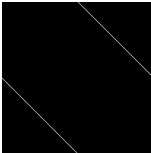
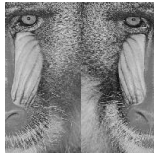
```
> A = imread('baboon_grey.jpg');
> A = double(A);
> imshow(uint8(2 * A));
> imshow(uint8(A));
> imwrite('baboon_rotated.jpg', uint8(A));
```

Note that we need to convert the image matrix to format double for matrix operations (such as transpose). For input and output, the matrix needs to be in format `uint8`.

Images Processing

 $A' =$  $2*A =$ 

Images Processing

 $B =$  $A*B =$ 

Processing Images

We can also multiply an image matrix with a matrix we have generated using Matlab:

```
> B = [eye(100) eye(100); eye(100) eye(100)];
> B = B - eye(200);
> imshow(uint8(255 * B));
> imshow(uint8(A * B));
```

We can convolve an image with a kernel using the `conv2` command (see next lecture for details):

```
> K = [1/9 1/9 1/9; 1/9 1/9 1/9; 1/9 1/9 1/9];
> C = conv2(K, A);
> imshow(uint8(C));
> K = [1 0 -1; 2 0 -2; 1 0 -1];
> D = conv2(K, A);
> imshow(uint8(D));
```

Example: Image Processing

 $C =$  $D =$ 

Summary

- Inverse: `inv(A)`;
- determinant: `det(A)`;
- eigenvalues and eigenvectors: `eig(A)`;
- plotting functions: `plot`, `fplot`;
- plotting discrete data: `bar`, `stem`;
- processing images: `imread`, `imwrite`, `imshow`;
- convolution: `conv2`.