## CogSci 109: Lecture 11

## Monday Oct 29, 2007

Return to classes - changes in the course plan, basic fits - regression, linear least squares

## Outline for today

■ Announcements
$\square$ Addressing the devastating fires

- How is our plan changing?
- How is our plan staying the same?
- Outline for today
$\square$ Reminder custom colormap demo
$\square$ Basic data fits
- Least squares minimization
- Linear models and regression
$\square$ Introduction
$\square$ Examples
$\square$ Matlab implementation


## Announcements

- Wildfires in San Diego
- Hw 3 due date is today, but it was going to be due Monday of last week, so you should at least have been mostly done
$\square$ If you have special needs in terms of time, come speak with me after class or in office hours
$\square$ If you are considering dropping, please discuss it with me first
- Reading for least squares and other fits


## Update: the big picture

- Where we are
$\square 4$ parts of the course
- We discussed data
$\square$ What is it, how do we manipulate it, matlab implementation
$\square$ Filtering
$\square$ Computing basic statistics
- We discussed basic visualization
$\square$ Plotting data (2d, 3d, colormaps)


## Update: the big picture (II)

- Where we're going
- We will now cover


## Modeling

- what is modeling?

Error analysis

- How good is your model?


## Update: the big picture (III)

- Where we're going (continued)
$\square$ What we're going to cover
- Basic models
$\square$ Linear fits, nonlinear fits
$\square$ Regression
$\square$ Relationship to machine learning
$\square$ Interpolation/extrapolation (also data analysis methods)
- Advanced models and modeling methods
$\square$ Fitting models with optimization methods
$\square$ Artificial neural networks
$\square$ AI
- Communicating results
$\square$ This has been integrated and will continue to be integrated
$\square$ Proper forms of inserting figures and tables in scientific communications
$\square$ Format in homeworks is designed to teach proper communication methodology


## Creating color maps - review and expansion

- What if I want to examine the boundaries of my data?
$\square$ I only want to see the extremes
$\square$ We can create a custom color map!



## Creating the color map ( $\mathbf{r}, \mathrm{g}, \mathrm{b}$ ) components

- To create a custom color map we need to make a matrix which is Dim nx3, range [0,1]
- Each column is the range of either red, green, blue
- Writing it by hand:

$$
M=\left[\begin{array}{lll}
0 & 1 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0 \\
1 & 0 & 0
\end{array}\right]
$$

- Typing it into a matlab variable:

$$
M=\left[\begin{array}{ccccc}
1 & 0 & 0 ; 0 & 0 & 0 \\
0 & 0 & 0 ; 0 & 0 & 0 ; \\
0 & 1 & 0
\end{array}\right]
$$

## Now what?

- We create our plot, let's create some data: X = peaks(50);
- And plot it using pcolor: pcolor(X) colormap(M)


## Here's what we get...

- As you can see this can be very useful for feature detection
■ But let's say we want to make a smooth map, how do we do that?



## Creating smooth color map

## functions

■ Instead of typing the matrix in manually, let's construct the functions we need to make transitions smooth from one color to the next

- Create many values in between 0 and 1
- Two things of note
$\square$ The length of your colormap array is up to you, the more numbers and the smaller the transitions, the more smooth the colors look (crayons vs. airbrushing)
$\square$ The colors are mapped so the range( 0,1 ) -> range( min(data), max(data) )


## Looking at smooth transitions

- Comparison after matching the number of values in the simple color variation (1-> 0 ) vs. a smooth function from 1->0
- Uses the equation...
$\square$ (for Decreasing:)

$$
\begin{aligned}
& r=\exp (-x) \\
& x=0: .01: 10
\end{aligned}
$$



## The final smooth color map

- And equations:

Decreasing:
$r=\exp (-x)$
$x=0: .01: 10$

Increasing:
$g=\frac{\exp (x)}{\max [\exp (x)]}$
$x=0: .01: 10$


## Results of our custom color

## map



## Other plots vs. custom color

## maps

- Grayscale?



■ Compressed?



## Matlab implementation...

■ To matlab...

## Part III: Models and the modeling process

## Linear least squares

- You're probably all familiar with linear regression -- fitting a line to a bunch of data.
- more formally fitting $\boldsymbol{y}=\boldsymbol{m} \boldsymbol{x}+\boldsymbol{b}$ for paired $\mathrm{x}, \mathrm{y}$ data (can also do multidimensional)
- Let's see how it's done mathematically


## Let's start by considering an easier question...

■ We have 2 points, and want to fit a line to them

- $(1,2),(3,4)$

■ How would you solve this problem?

- We want $\mathrm{y}=\mathrm{mx}+\mathrm{b}$ (we need $\mathbf{m}$ and $\mathbf{b}$ )
$\square$ Substitute each point in

$$
\begin{aligned}
& 2=\mathrm{m}(1)+\mathrm{b} \\
& 4=\mathrm{m}(3)+\mathrm{b}
\end{aligned}
$$

## Example continued

- And solve for $\mathbf{b}$ first, then $\mathbf{m}$
$\mathrm{b}=2-\mathrm{m}$
$4=3 \mathrm{~m}+2-\mathrm{m}$
$4=2 \mathrm{~m}+2$
$\mathrm{~m}=1$
$\mathrm{~b}=2-\mathrm{m}$
$\mathrm{b}=1$


## Example continued

■ We have two equations and two unknowns ( $\mathrm{m}, \mathrm{b}$ )

- This can be written compactly as

$$
\left[\begin{array}{ll}
1 & 1 \\
3 & 1
\end{array}\right]\left[\begin{array}{l}
m \\
b
\end{array}\right\}=\left\{\begin{array}{l}
2 \\
4
\end{array}\right\}
$$

- Which is of the basic form

$$
A x=b
$$

- We want to find

$$
x=A^{-1} b
$$

## Solving Ax=b

- Solving for $x=A^{-1} b$ involves computing the inverse of the A matrix
$\square$ Insiwhatsitz? Don't worry...inverses are a way to make life easier
- There are several methods, and you can solve for arbitrarily sized problems (ie what if we want to find 100 variables? Not fun by hand:( Let's use a computer to do it for us!!!:)
$\square$ Gaussian elimination (what you learned in linear algebra class)
- Don't worry you won't have to do it by hand in this class!
$\square$ Thomas algorithm, etc (and other more efficient methods computationally)
$\square$ Matlab has gaussian elimination built-in nicely of course


## We need to remind ourselves of matrix inversion

- What is an inverse of a matrix?
- Rotation example
$\square$ If a vector is rotated by multiplying it by a rotation matrix, then multiplying the rotated vector by the inverse rotates the vector back to its original orientation
$\square$ Side note - a matrix times its inverse yields the identity matrix
- You can test for a matrix being the inverse of another matrix by multiplying the two, and see how close do you get to the identity matrix?

$$
\begin{aligned}
& A A^{-1}=I \\
& \text { Look up more of the definition details...see references on site }
\end{aligned}
$$

- Homework problem, one matrix plot is an example...which could it be? Hmm... what special matrices have we just mentioned? Hmmm...how could I IDENTIFY this matrix? Hmmm...
- Dating example


## Solving Ax=b

■ We compute the solution of our canonical problem by

|  | $\begin{array}{r} \text { Recall } \\ \text { that... } \end{array}$ | $A x=b$ |
| :---: | :---: | :---: |
| $A^{-1} A=I$ |  | $A^{-1} A x=A^{-1} b$ |
| $I x=x$ |  | $I x=A^{-1} b$ |
|  |  | $x=A^{-1} b$ |

## How to solve $\mathbf{A x}=b$ in matlab

- In matlab this can be solved for with the $\backslash$ operator
- $\mathrm{A} \backslash \mathrm{B}$ is the matrix division of B into A
$\square$ roughly the same as INV(A)*B
$\square$ computed in a different way.
- If A is an N -by- N matrix and B is a column vector with N components, or a matrix with several such columns, then $X=A \backslash B$ is the solution to the equation $\mathrm{A} * \mathrm{X}=\mathrm{B}$ computed by Gaussian elimination.
- Doing it in matlab:

$$
m b=[11 ; 31] \backslash[2 ; 4] ; \quad \%(\text { left matrix divide })
$$

## Derivation of linear least

squares

- <on board>


## Another example in matlab

- consider $(1,2)(3,4)(2,3.5)$
$\mathrm{x}=\left[\begin{array}{lll}1 & 3 & 2\end{array}\right]$
$y=\left[\begin{array}{lll}2 & 4 & 3.5\end{array}\right]$
plot(x,y,'*')
- $1 \mathrm{~m}+\mathrm{b}=2$
- $3 \mathrm{~m}+\mathrm{b}=4$
- $2 \mathrm{~m}+\mathrm{b}=3.5$


## Example continued

$$
\begin{aligned}
& A=[11 ; 31 ; 21] \\
& y=[2 ; 4 ; 3.5]
\end{aligned}
$$

- if we use the $m=1, b=1$ solution to the first two it doesn't fit the third
- e.g. 3 equations and 2 unknowns
- This is what is known as an overconstrained problem. People commonly like to find the solution that minimizes the mean square error


## Example continued

- This means we want to find the solution that minimizes
Isum_\{(x,y) pairs\} (y-mx-b)^2
- Matlab again solves this with

```
mb=A\y
hold on
newA=[00 1; 5 1]
plot([0 5],newA*mb)
```

