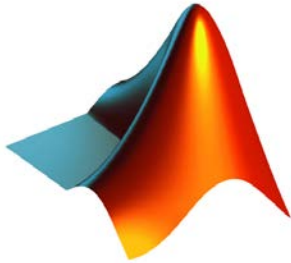
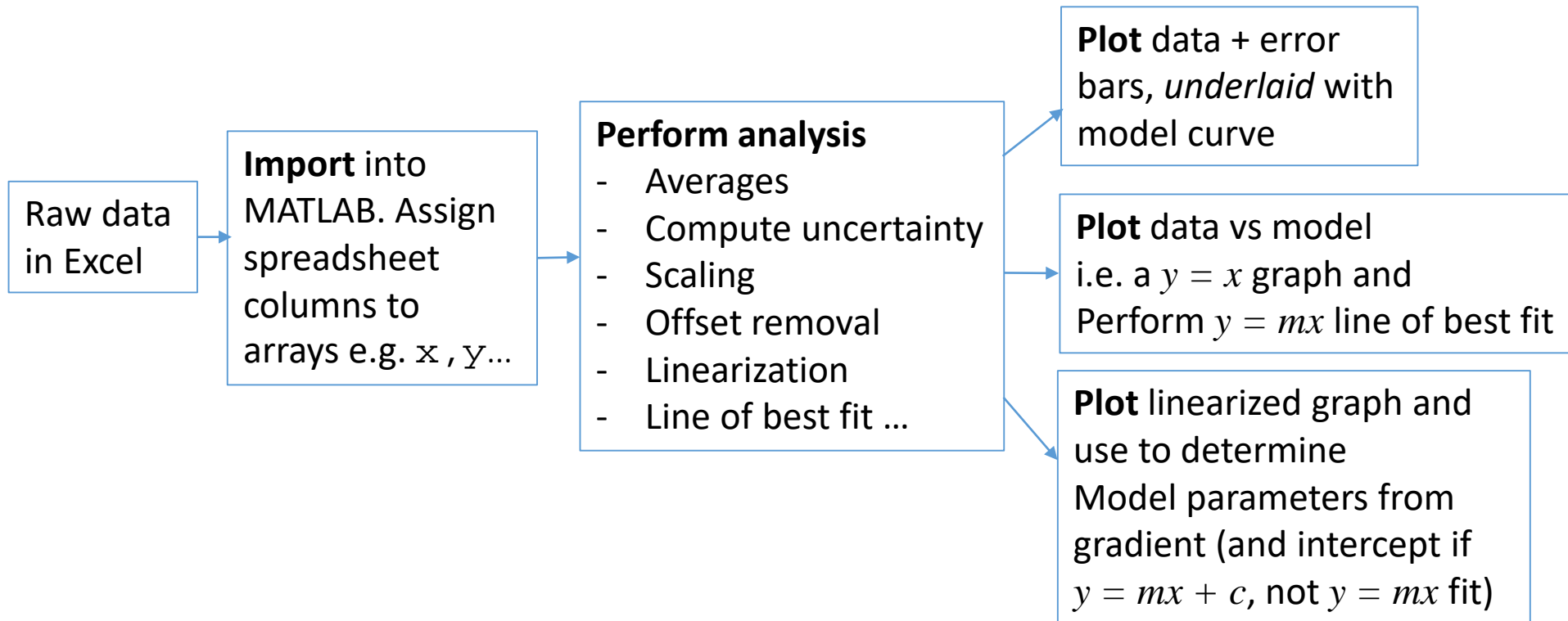


# Experimental data processing pipeline using MATLAB



Dr Andrew French. July 2020.



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# Physics Revision Notes

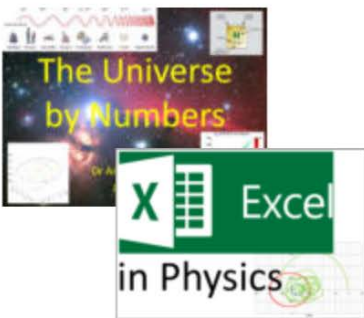
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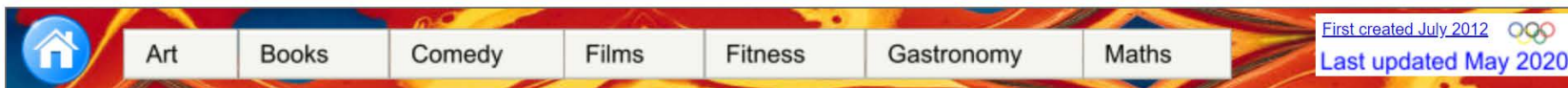
Feedback stamps  
for teachers



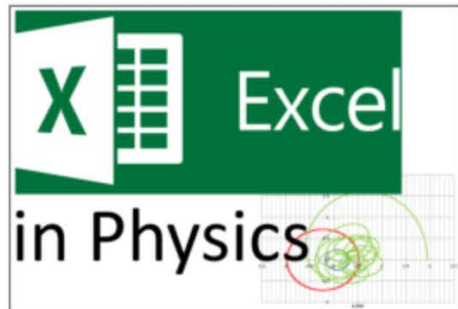
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► Intro

[Project vs skills matrix](#)

## Simple calculators

**SPEED CONVERTER**

SPEED IN METRES PER SECOND  
20

SPEED / KM PER HOUR 72.00

SPEED / MILES PER HOUR 44.74

SPEED / NAUTICAL MILES PER HOUR 38.88

SPEED / NANOPARSEC PER BEARD-INCH 2.94

speed = distance / time  
mile = 1609m  
nmile = 1852m  
ms<sup>-1</sup> = 2.237mph  
ms<sup>-1</sup> = 3.600kmh<sup>-1</sup>  
parsec = 3.086 × 10<sup>16</sup>m  
beard-inch = 7.5 weeks

nanoparsec =  $\frac{10^{-9} \times 3.086 \times 10^{16} \text{ m}}{7.5 \times 7 \times 24 \times 3600 \text{ s}} = 6.80 \text{ ms}^{-1}$

[Speed converter](#)  
[Snell's law of refraction](#)  
[Radio frequency & wavelength](#)

## Maths tools

**ERROR CALCULATION**

ACTUAL X VALUE 122

X VALUES WITH RANDOM ERROR

121	121	125	122	120	128	120	121	124	119	N
										10

MEAN X  $\bar{x} = \frac{1}{N} \sum_{i=1}^N x_i$   
122

$(x_i - \bar{x})^2$

1.21	1.21	8.41	0.01	4.41	34.8	4.41	1.21	3.61	9.61
------	------	------	------	------	------	------	------	------	------

ERROR IN X  $\sigma_x = \sqrt{\frac{1}{N-1} \sum_{i=1}^N (x_i - \bar{x})^2}$   
3

SD:  $s = (122 \pm 3)$

► Error analysis  
[Errors Excel sheet](#)

## Excel techniques:

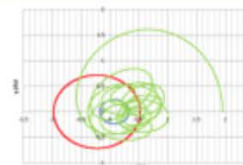
How to .... videos  
Number & text formatting  
[Formulae & replication](#)  
[Using \\$ prefix constants](#)  
[Plotting scatter graphs](#)  
[Underlying model curves](#)  
[Lines of best fit](#)  
[Plotting error bars](#)  
IF statements in formulae  
[Printing to PDFs and PNGs](#)  
[Dealing with lots of data](#)  
[Solar system Excel sheet](#)

## Virtual experiments

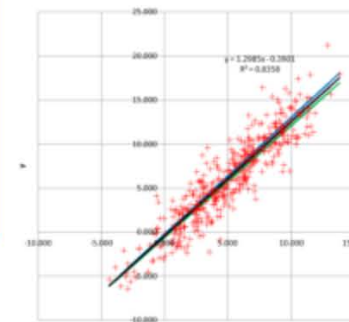
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[Hubble's law](#)  
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►► [Capacitor discharge](#)

## Simulations

[1D gravity + drag](#)  
[Epidemics](#)  
[2D projectiles + drag](#)  
[1D water rocket](#)  
[Orbits & galaxies](#)



## ► Lines of best fit



[Lines of best fit Excel sheets & notes](#)

⏪ A Course in Coding

All the downloads are .zip files. To edit the Excel sheets you will need to extract them to one of your own personal folders

[Download the whole course \(85.1MB\)](#)

Mountaineering

Music

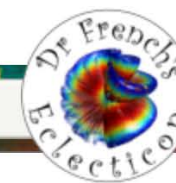
Philosophy

Photography

Physics

Programming

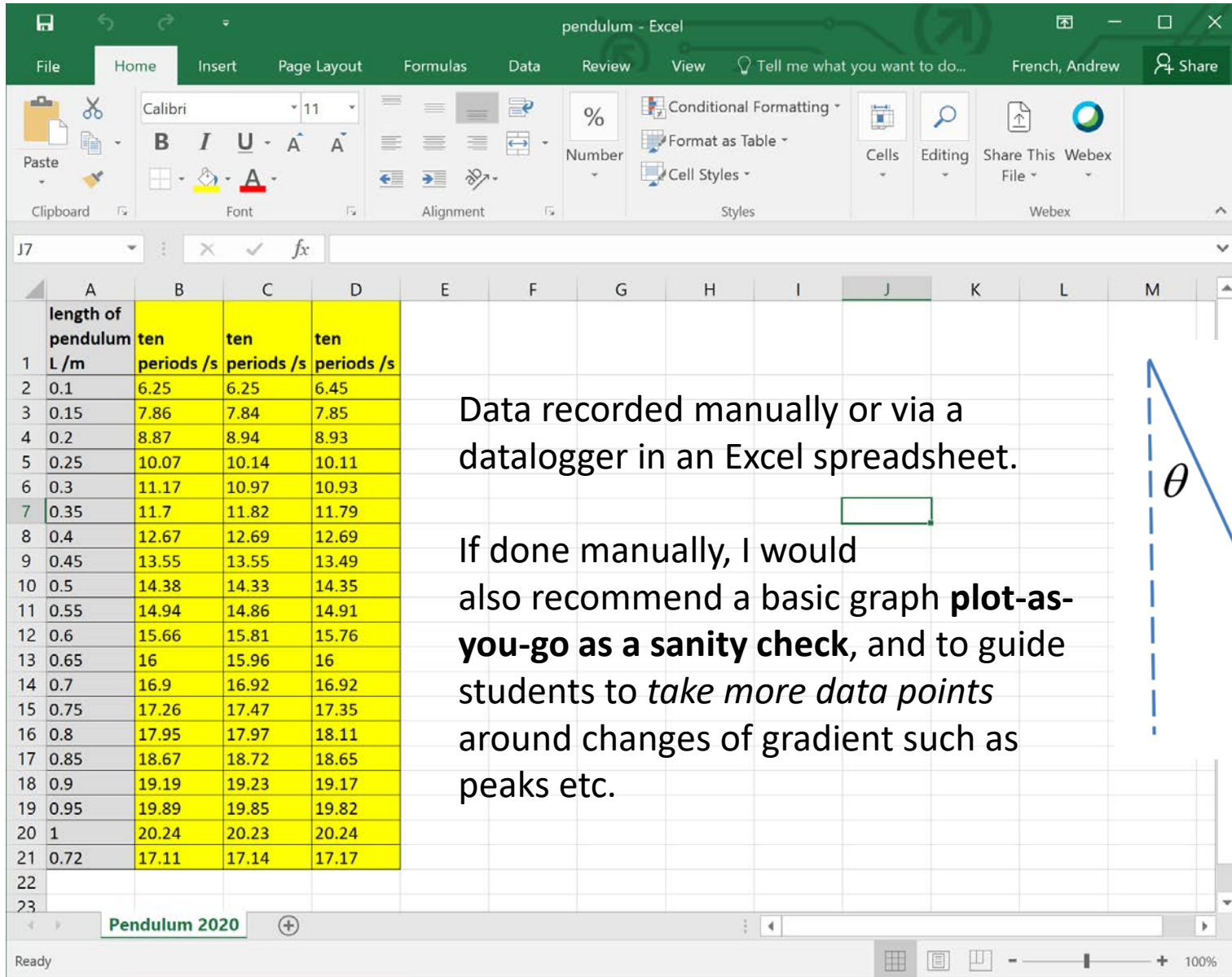
Writing





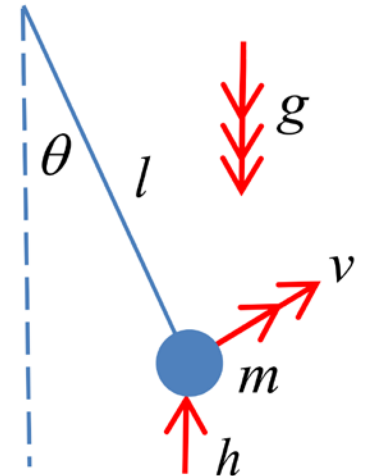
# Example using pendulum data

Raw data  
in Excel



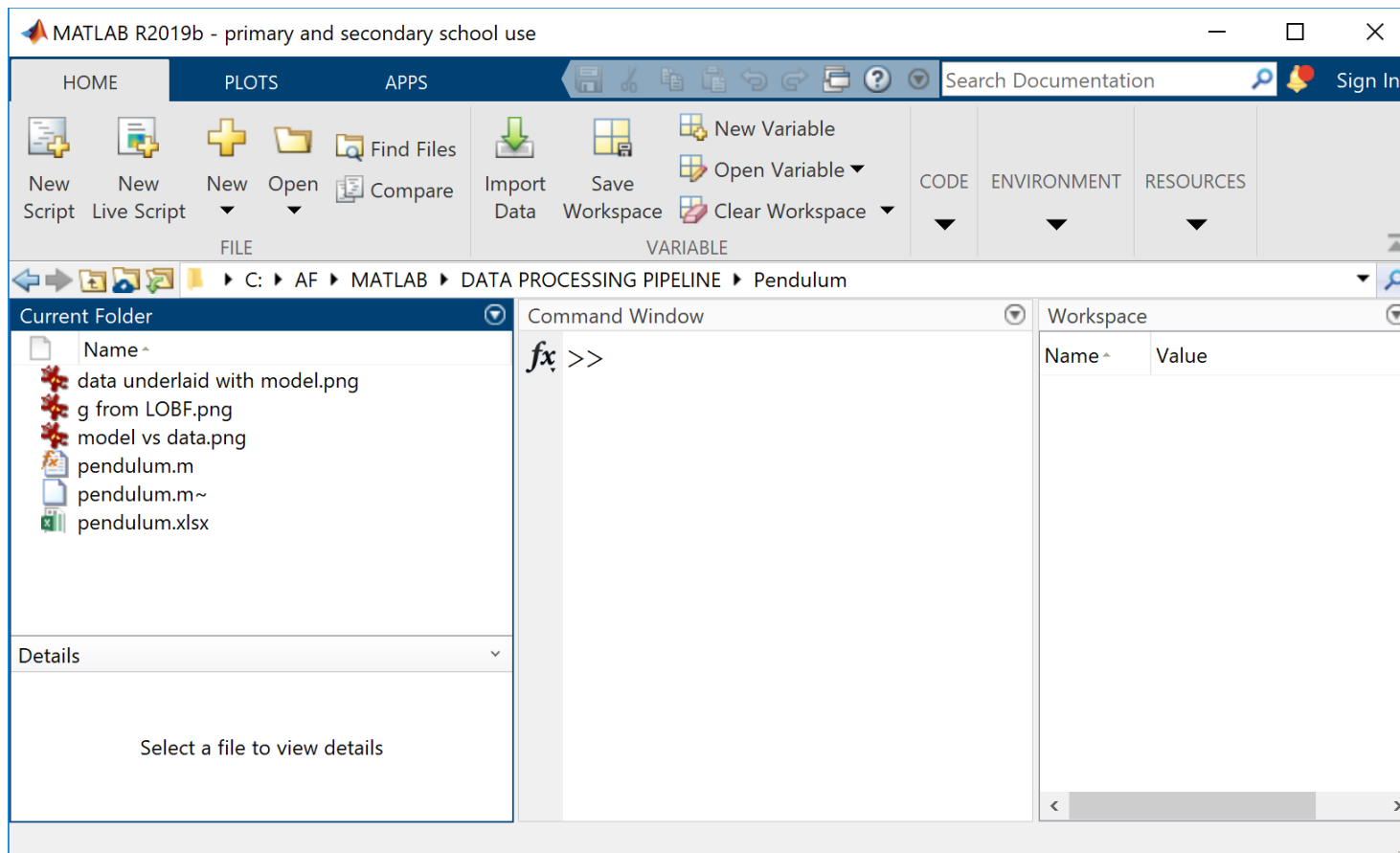
Data recorded manually or via a datalogger in an Excel spreadsheet.

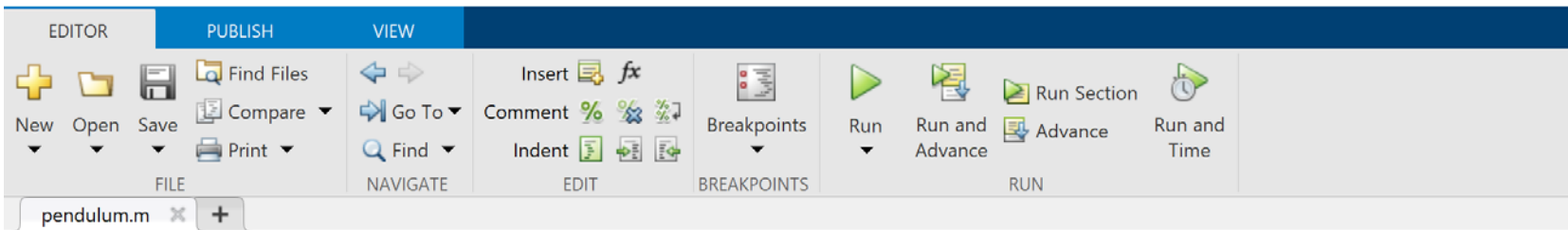
If done manually, I would also recommend a basic graph **plot-as-you-go** as a **sanity check**, and to guide students to *take more data points* around changes of gradient such as peaks etc.



Run `pendulum.m` (right click, **run**) to execute a series of commands which constitute the rest of the data processing pipeline. The code can be modified for different experiments.

The key feature is that the code performs the process *automatically*, which can save considerable time when working on new data sets. MATLAB has the ability to perform useful analysis and create bespoke plots to a much higher standard than Excel. Students can focus on the *process*, in modifying the code, rather than the faff of dealing with Excel's defaults! However, I would always start with Excel as a first IT-based analysis.





```

1 % Example physics data processing pipeline: #1 Pendulum
2 % * Load raw data from an Excel sheet pendulum.xlsx. This has columns of
3 % pendulum length L /m, and three repeats of ten periods (10*T) /s.
4 % * Determine averages and errors
5 % * Plot  $y = 4\pi^2 * L$  vs  $x = T^2$ . Determine line of best fit (LOBF) and error,
6 % and hence determine g from data. Compare to  $g = 9.81\text{N/kg}$ .
7 % * Plot T (data) vs  $2\pi\sqrt{L/g}$  (with actual g). Perform LOBF.
8 % * Underlay T vs L data and underlay with  $T = 2\pi\sqrt{L/g}$  model.
9 %
10 % LAST UPDATED by Dr Andrew French. July 2020.

```

## Inside `pendulum.m`

.... It is a text file!

`%` means **commentary**

- Vital for humans
- Ignored by machines

```

12 function pendulum
13
14 %% INPUTS %%
15
16 %Fontsize and marker size for graphs
17 fsize = 18; msize = 18;
18
19 %Set (fixed) error (in m) for pendulum length. Assume no systematic error.
20 Lerror = 0.01;
21
22 %Actual value of g /Nkg^-1
23 g = 9.81;
24
25 %Leave figures or close after printing?
26 close_after_print = 1;

```

length of pendulum	ten periods /s	ten periods /s	ten periods /s
L /m	ten periods /s	ten periods /s	ten periods /s
0.1	6.25	6.25	6.45
0.15	7.86	7.84	7.85
0.2	8.87	8.94	8.93
0.25	10.07	10.14	10.11
0.3	11.17	10.97	10.93
0.35	11.7	11.82	11.79
0.4	12.67	12.69	12.69
0.45	13.55	13.55	13.49
0.5	14.38	14.33	14.35
0.55	14.94	14.86	14.91
0.6	15.66	15.81	15.76
0.65	16	15.96	16
0.7	16.9	16.92	16.92
0.75	17.26	17.47	17.35
0.8	17.95	17.97	18.11
0.85	18.67	18.72	18.65
0.9	19.19	19.23	19.17
0.95	19.89	19.85	19.82
1	20.24	20.23	20.24
0.72	17.11	17.14	17.17

**Import into MATLAB. Assign spreadsheet columns to arrays e.g. x, y...**

29

30 `%% IMPORT EXCEL DATA & PREPARE L, T arrays %%`

31

32 `%Import data. Four columns. First is pendulum length, next three are`  
 33 `% ten periods /s.`

34 `[num,txt,row] = xlsread( 'pendulum' );`

35 `L = num(:,1); T10_1 = num(:,2); T10_2 = num(:,3); T10_3 = num(:,4);`

36

37 `%Determine period T /s and the (unbiased estimator) of the error in T.`

38 `%The second argument of the std function uses the /(N-1) normalization`

39 `T = mean( [T10_1 , T10_2 , T10_3 ],2 )/10;`

40 `Error = std( [T10_1 , T10_2 , T10_3 ],0,2 )/10;`

```

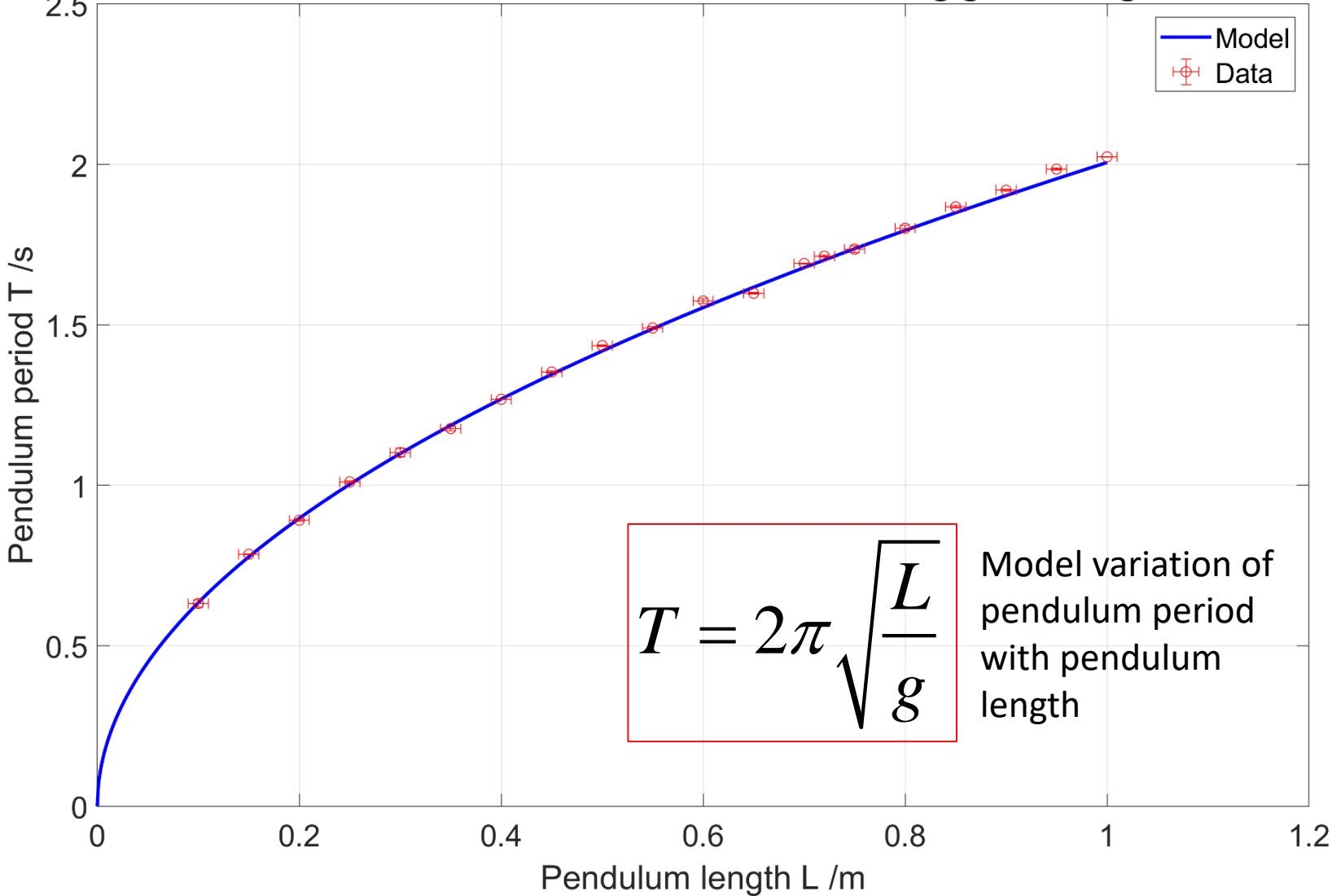
44 %% ANALYSIS: Compare T vs L data to model T(L) with actual g %%
45
46 %Determine model prediction of T using actual value of g
47 Tmodel = 2*pi*sqrt( L/g );
48
49 %Determine model at a much finer grid of L values
50 LL = linspace( 0,max(L),1000 ); TTmodel = 2*pi*sqrt( LL/g );
51
52 %Plot model curve of T vs L
53 figure('name','model vs data','color',[1 1 1],...
54         'units','normalized','position',[0.05, 0.05, 0.9, 0.85]);
55 plot( LL, TTmodel,'b-','linewidth',2 ); hold on;
56 set( gca, 'fontsize',fsize ); grid on;
57
58 %Plot data error bars
59 x = L; y = T; yneg = Terror; ypos = Terror;
60 xneg = Lerror*ones(size(L)); xpos = Lerror*ones(size(L));
61 errorbar( x,y,yneg,ypos,xneg,xpos,'o','color','r');
62
63 %Graph labels etc
64 xlabel('Pendulum length L /m'); ylabel('Pendulum period T /s');
65 title('Pendulum data underlaid with model using g=9.81N/kg');
66 legend({'Model','Data'});
67
68 %Print a PNG file
69 print((gcf,'data underlaid with model.png','-r300','-dpng') );
70 if close_after_print==1; close(gcf); end

```



**Plot data + error bars, *underlaid* with model curve**

**Pendulum data underlaid with model using  $g=9.81\text{N/kg}$**



```

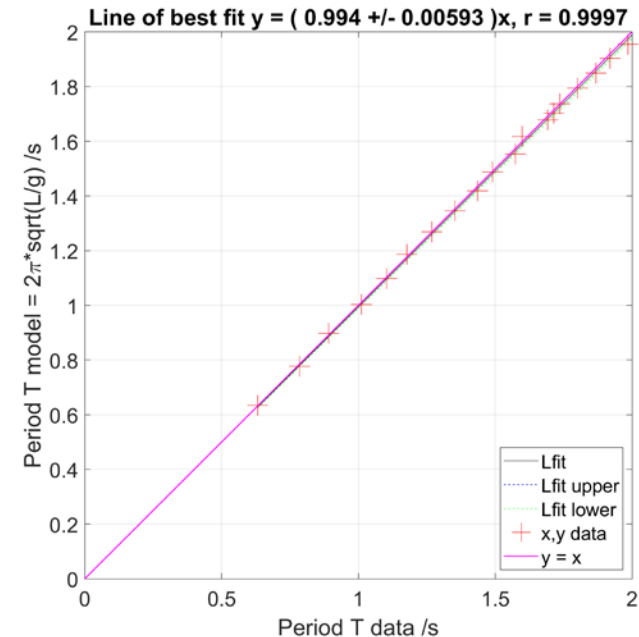
74 %% ANALYSIS: Determine line of best fit of the form y = m*x between T data and T model
75 % For 100% correlation, the gradient m = 1 and product-moment correlation coefficient r = 1.
76 y = Tmodel; x = T; [yfit,xfit,r,m,dm,yupper,ylower,s] = bestfit( x,y );
77
78 %Plot line of best fit
79 xlabel_str = 'Period T data /s';
80 ylabel_str = 'Period T model = 2\pi*sqrt(L/g) /s';
81 plot_LOBF( x,y, yfit,xfit,r,m,dm,yupper,ylower,...
82     fsize, msize, xlabel_str, ylabel_str );
83
84 %Plot y = x for visual check
85 plot( [0;x], [0;x], 'm-', 'linewidth', 1 );
86 legend({'Lfit', 'Lfit upper','Lfit lower','x,y data','y = x'},...
87     'location','southeast'); axis equal; axis tight;
88
89 %Set sensible x,y limits to include origin
90 xlims = get(gca,'xlim'); set( gca, 'xlim',[0,round( xlims(2) )] );
91 ylims = get(gca,'ylim'); set( gca, 'ylim',[0,round( ylims(2) )] );
92 print((gcf, 'model vs data.png','-r300','-dpng' );
93 if close_after_print==1; close(gcf); end
94

```

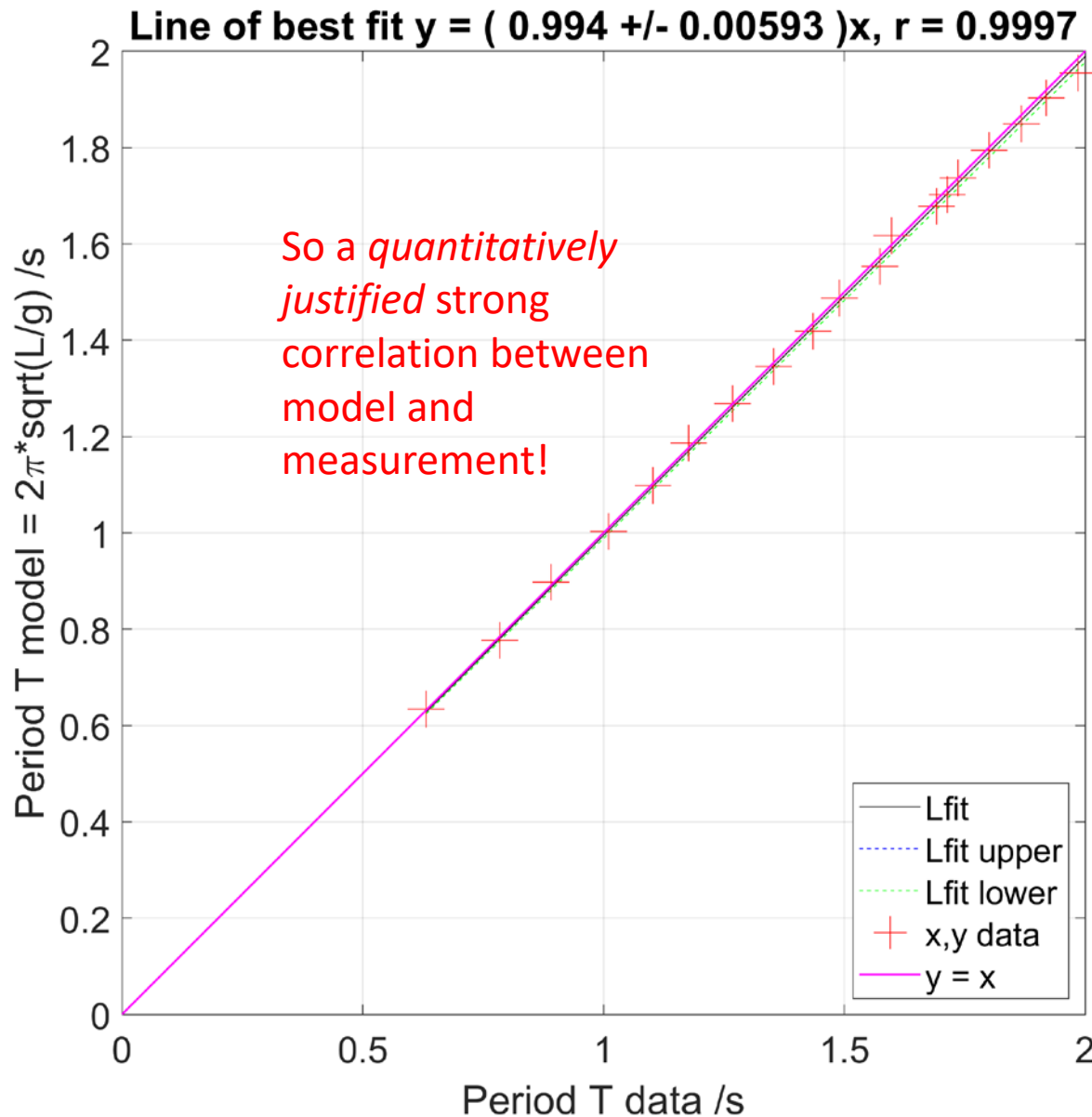
These are *sub-functions* which perform the line of best fit and associated plots. They should be generic, regardless of the dataset.

$$T = 2\pi \sqrt{\frac{L}{g}}$$

Model variation of pendulum period  $T$  with pendulum length  $L$



**Plot data vs model**  
i.e. a  $y = x$  graph and  
Perform  $y = mx$  line of best fit



$$T = 2\pi\sqrt{\frac{L}{g}}$$

Model variation of  
pendulum period  
with pendulum  
length

If you don't need to find  
parameters from data,  
**simply comparing model vs  
measurement** is a very clear  
first quantitative analysis

```
%% ANALYSIS: Determine g from data %%
```

```
%Determine  $y = 4\pi^2 L$  and  $x = T^2$ 
```

```
x = T.^2; y = 4*pi^2 * L;
```

```
%Determine upper and lower values for error bar calculation
```

```
x_upper = ( T + Terror ).^2; x_lower = ( T - Terror ).^2;
```

```
y_upper = 4*pi^2 * ( L + Lerror ); y_lower = 4*pi^2 * ( L - Lerror );
```

```
% Determine line of best fit of the form  $y = m*x$ .
```

```
% Gradient m is g in this case
```

```
[yfit,xfit,r,m,dm,yupper,ylower,s] = bestfit(x,y);
```

```
%Plot line of best fit
```

```
xlabel_str = '(T/s)^2'; ylabel_str = '4\pi^2*(L/m)';
```

```
plot_LOBF( x,y, yfit,xfit,r,m,dm,yupper,ylower,...
```

```
    fsize, 0.001, xlabel_str, ylabel_str );
```

```
%Plot what the line should be, given the actual value of g
```

```
plot( x, g*x, 'm-', 'linewidth',1 );
```

```
%Plot data error bars
```

```
yneg = y - y_lower; ypos = y_upper - y; xneg = x - x_lower; xpos = x_upper - x;
```

```
errorbar( x,y,yneg,ypos,xneg,xpos,'o','color','r');
```

```
%Add a legend
```

```
legend({'Lfit', 'Lfit upper','Lfit lower','','...'
```

```
    'Using g=9.81N/kg','x,y data'}, 'location','southeast' )
```

But if you *do* need to find parameters, **linearize**,  
and then perform a line of best fit

$$T = 2\pi \sqrt{\frac{L}{g}}$$
$$\therefore \underbrace{4\pi^2 L}_y = g \underbrace{T^2}_x$$
$$\Rightarrow y = gx$$

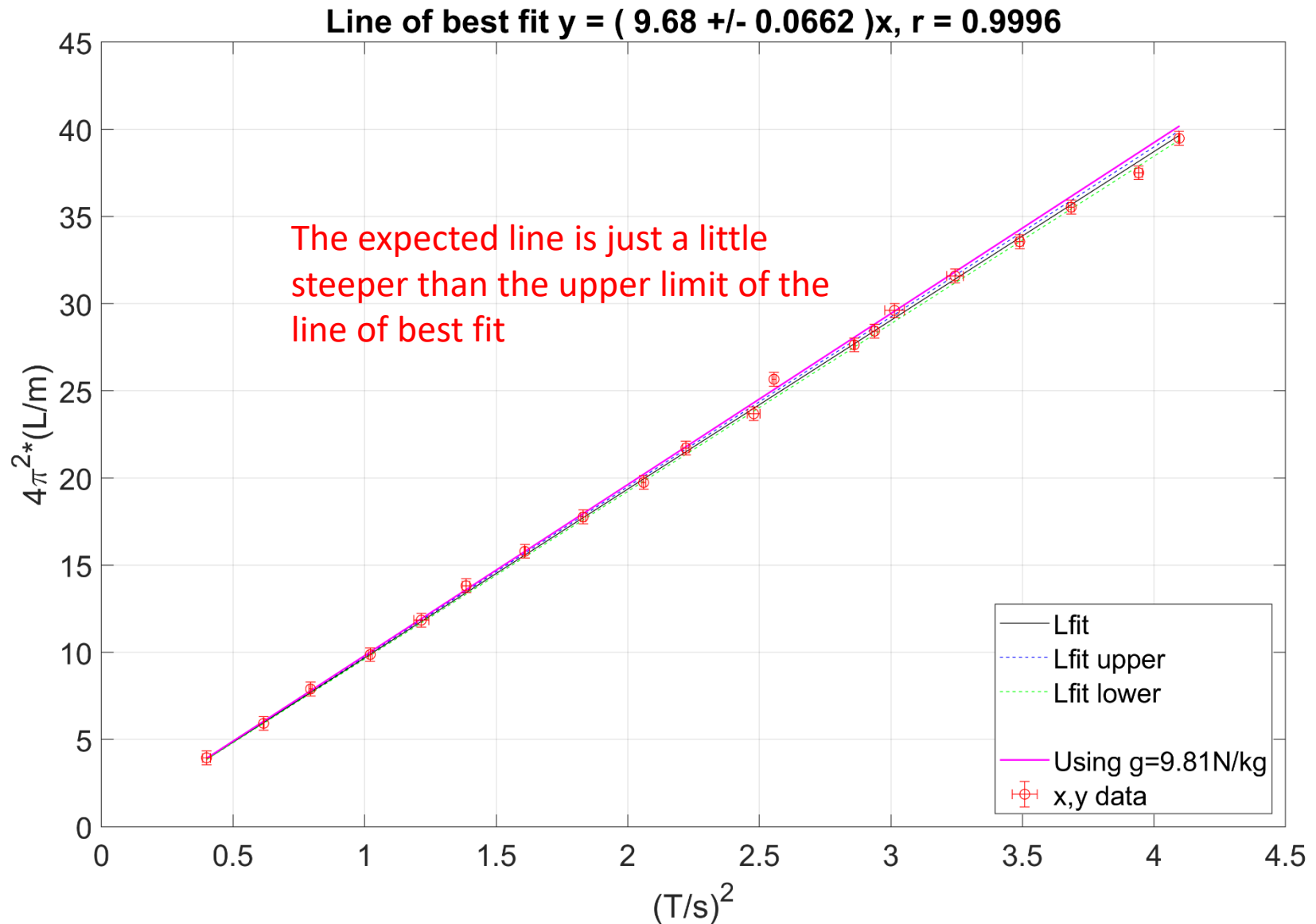
So  $g$  is the gradient of the  $x,y$  graph in our case

**Plot** linearized graph and  
use to determine model  
parameters from gradient  
(and intercept if  
 $y = mx + c$ , not a  $y = mx$  fit)



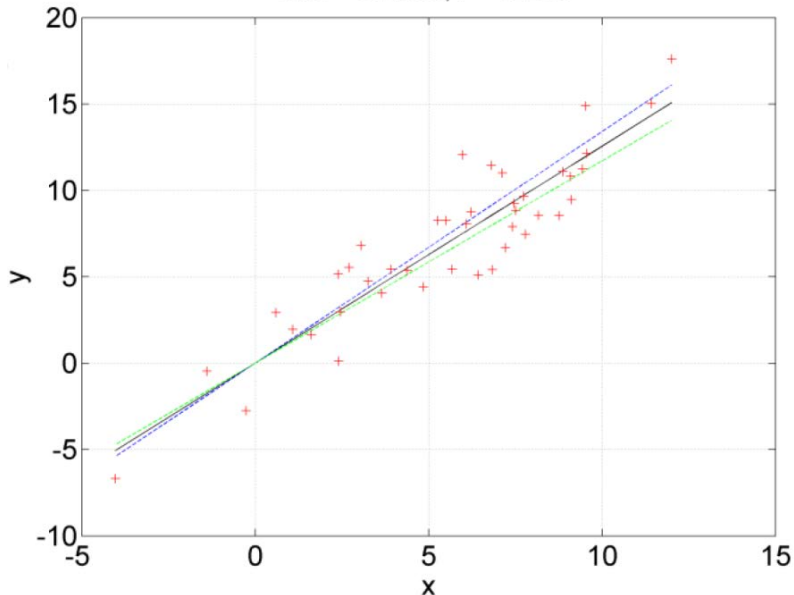
$$T = 2\pi \sqrt{\frac{L}{g}} \quad \therefore \underbrace{4\pi^2 L}_y = g \underbrace{T^2}_x \quad \Rightarrow y = gx$$

In our case, our gradient (and hence calculated  $g$ ) is systematically lower than what it should be.

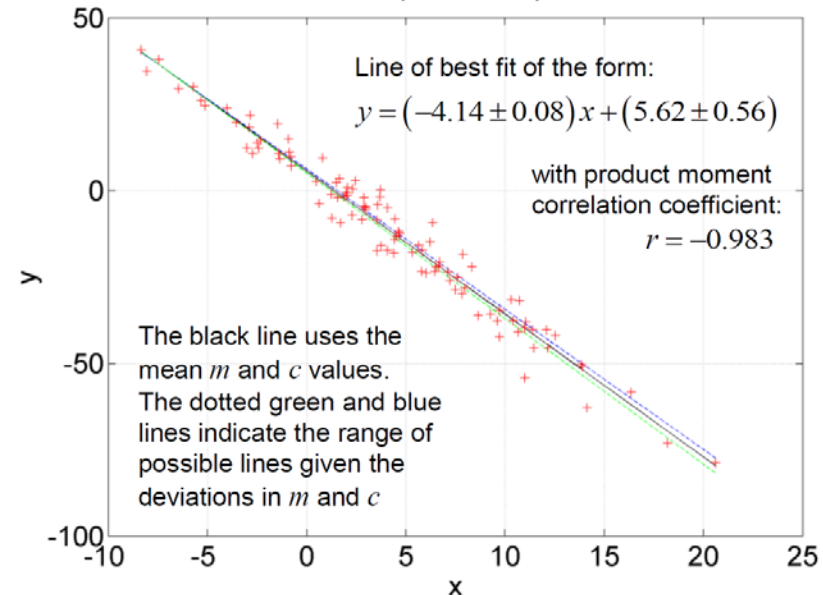


# Lines of best fit using MATLAB

Line of best fit  $y = 1.26x$   
 $\Delta m = 0.0853, r = 0.916$

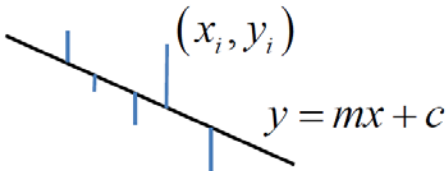


Line of best fit  $y = -4.14x + 5.62$   
 $\Delta m = 0.0783, \Delta c = 0.56, r = -0.983$

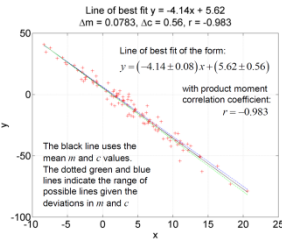


To determine the line of best fit\*, let us sum the *squared* deviations of  $(x,y)$  from the line of best fit.

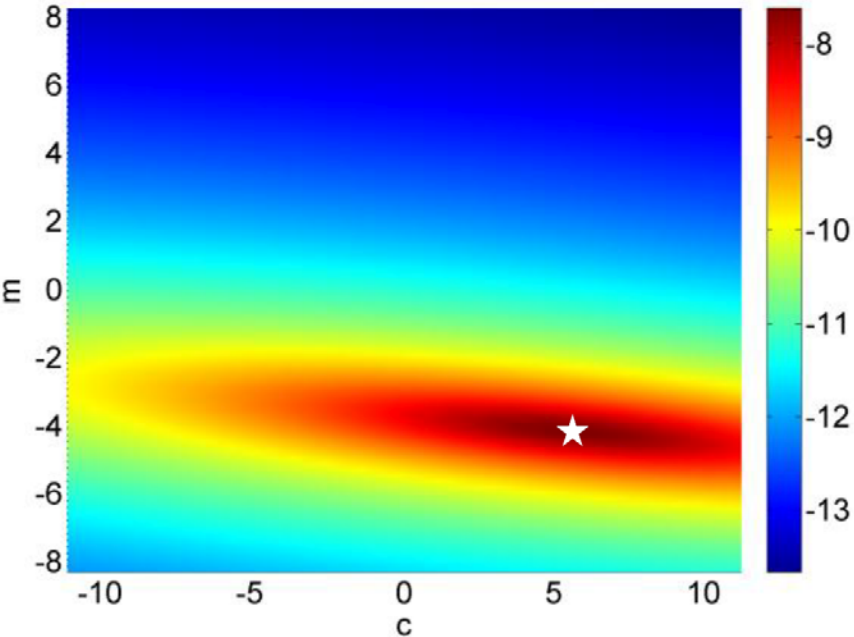
$$S = \sum_{i=1}^N (y_i - mx_i - c)^2$$



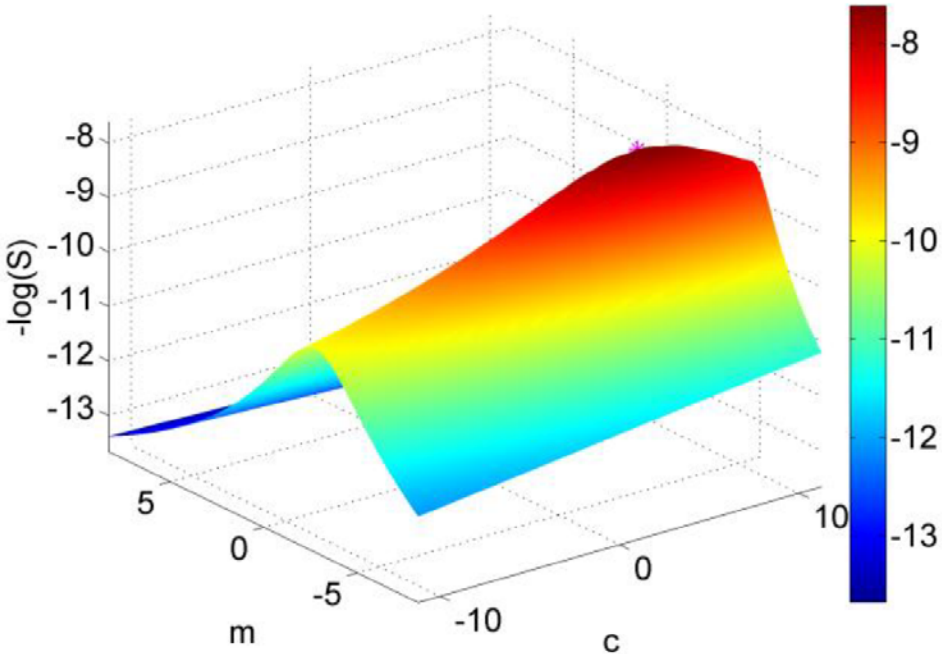
Using the (*negatively correlated*) data on the right, we can plot a surface of  $S$  vs  $m$  and  $c$  values. We can see this has a **minimum** at a particular  $(m,c)$  coordinate. (Note for clarity the plots below are of  $-\log S$ , so the  $(m,c)$  coordinate corresponds to the peak, i.e. maximum, instead).



$-\log(\text{Sum of } (y - mx - c)^2)$   
 $m = -4.14, c = 5.62$



$-\log(\text{Sum of } (y - mx - c)^2)$   
 $m = -4.14, c = 5.62$



The idea is to find optimum  $m$  and  $c$  values (or just  $m$ ) given  $x,y$  data

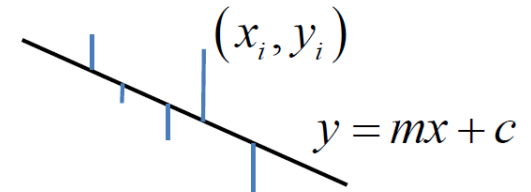
## Correlation & Linear Regression

Perhaps the most important analytical tool in the physical sciences is the ability to quantify the validity of a model relating a set of measurable parameters. The idea is as follows:

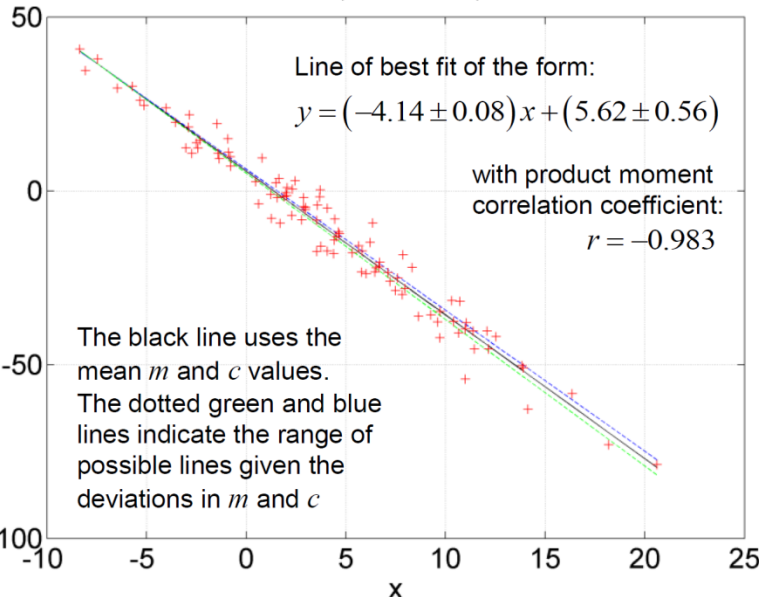
- (1) Rearrange the model in such a way that it becomes a *linear equation* of the form  $y = mx + c$
- (2) Plot experimental  $(x,y)$  data on a graph and determine the **line of best fit** through the data.
- (3) Determine *gradient*  $m$  and *vertical intercept*  $c$  from the line of best fit.
- (4) Determine the standard deviation of both gradient  $m$  and intercept  $c$ , and a quantitative measure of how good the fit is (this is called the **product moment correlation coefficient**).

To determine the line of best fit\*, let us sum the *squared* deviations of  $(x,y)$  from the line of best fit.

$$S = \sum_{i=1}^N (y_i - mx_i - c)^2$$



Line of best fit  $y = -4.14x + 5.62$   
 $\Delta m = 0.0783$ ,  $\Delta c = 0.56$ ,  $r = -0.983$



We can find  $S$  given a range of  $m$  and  $c$  values. Which pairing results in the *minimum* value of  $S$ ?



Summary: Line of Best Fit for:

$$y = mx + c$$

$N$  data point pairs  $(x, y)$

$$\bar{x} = \frac{1}{N} \sum_{i=1}^N x_i, \quad \bar{y} = \frac{1}{N} \sum_{i=1}^N y_i, \quad \overline{x^2} = \frac{1}{N} \sum_{i=1}^N x_i^2, \quad \overline{y^2} = \frac{1}{N} \sum_{i=1}^N y_i^2, \quad \overline{xy} = \frac{1}{N} \sum_{i=1}^N x_i y_i$$

$$V[x] = \overline{x^2} - \bar{x}^2, \quad V[y] = \overline{y^2} - \bar{y}^2, \quad \text{cov}[x, y] = \overline{xy} - \bar{x}\bar{y}$$

$$m = \frac{\overline{xy} - \bar{x}\bar{y}}{\overline{x^2} - \bar{x}^2} = \frac{\text{cov}[x, y]}{V[x]}, \quad c = \bar{y} - m\bar{x}$$

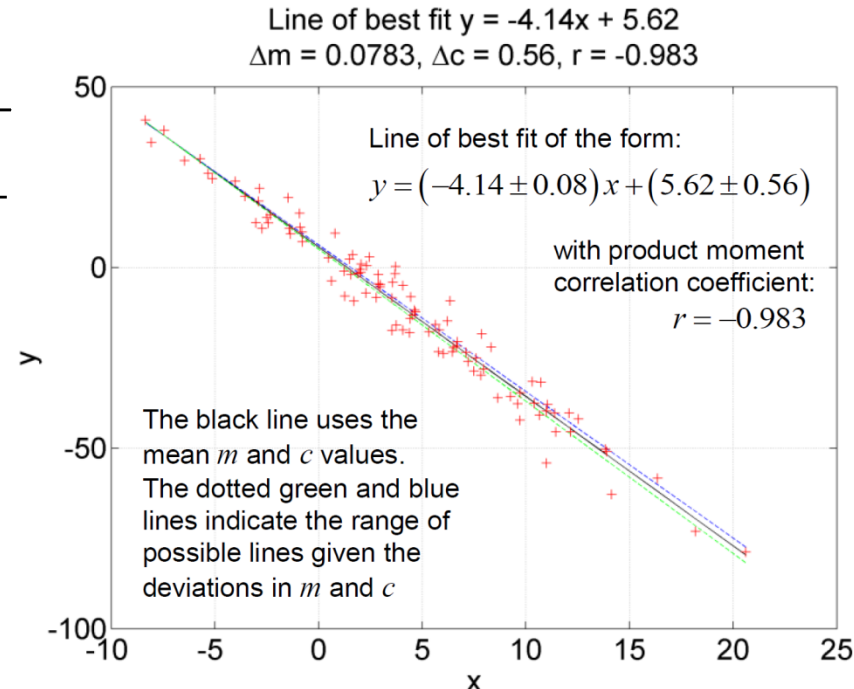
$$r = \frac{\text{cov}[x, y]}{\sqrt{V[x]V[y]}}$$

Product moment correlation coefficient

$$\Delta m = \frac{s}{\sqrt{N}} \frac{1}{\sqrt{V[x]}}, \quad \Delta c = \frac{s}{\sqrt{N}} \sqrt{1 + \frac{\bar{x}^2}{V[x]}}$$

$$s = \sqrt{\frac{1}{N-2} \sum_{i=1}^N (y_i - mx_i - c)^2}$$

Errors in gradient and intercept



Summary: Line of Best Fit for:

$$y = mx$$

$N$  data point pairs  $(x, y)$

$$\bar{x} = \frac{1}{N} \sum_{i=1}^N x_i, \quad \bar{y} = \frac{1}{N} \sum_{i=1}^N y_i, \quad \overline{x^2} = \frac{1}{N} \sum_{i=1}^N x_i^2, \quad \overline{y^2} = \frac{1}{N} \sum_{i=1}^N y_i^2, \quad \overline{xy} = \frac{1}{N} \sum_{i=1}^N x_i y_i$$

$$V[x] = \overline{x^2} - \bar{x}^2, \quad V[y] = \overline{y^2} - \bar{y}^2, \quad \text{cov}[x, y] = \overline{xy} - \bar{x}\bar{y}$$

$$m = \frac{\overline{xy}}{\overline{x^2}}$$

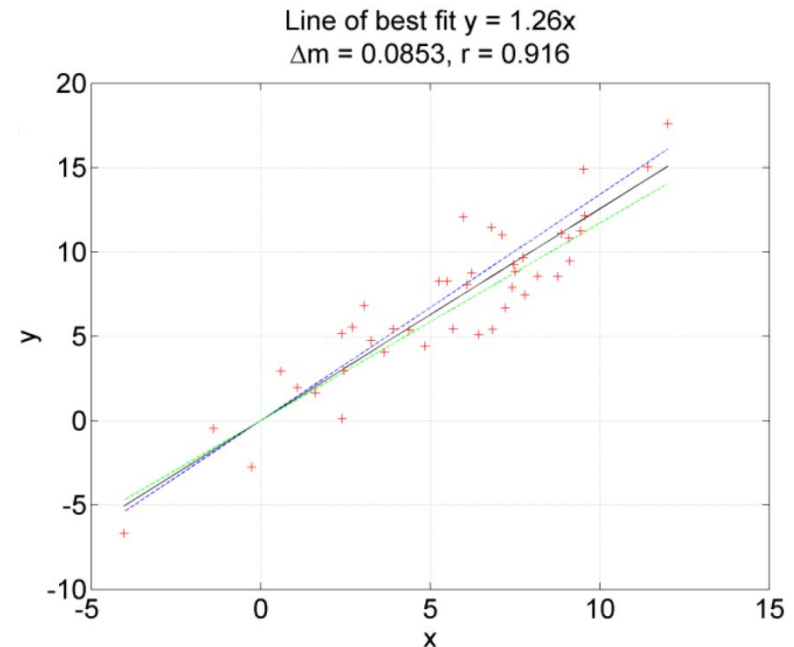
$$r = \frac{\text{cov}[x, y]}{\sqrt{V[x]V[y]}}$$

Product moment  
correlation coefficient

$$\Delta m = \frac{s}{\sqrt{N}} \frac{1}{\sqrt{V[x]}}$$

Error in gradient

$$s = \sqrt{\frac{1}{N-1} \sum_{i=1}^N (y_i - mx_i)^2}$$



```
%Line of best fit function  $y_{fit} = m*x$ , with product moment correlation  
%coefficient  $r$ 
```

```
function [yfit,xfit,r,m,dm,yupper,ylower,s] = bestfit(x,y)
```

```
%Find any x or y values that are NaN or Inf
```

```
ignore = isnan(abs(x)) | isnan(abs(y)) | isinf(abs(x)) | isinf(abs(y));
```

```
x(ignore) = []; y(ignore) = [];
```

```
%Compute line of best fit
```

```
xbar = mean(x); ybar = mean(y); xybar = mean(x.*y);
```

```
xxbar = mean(x.^2); yybar = mean(y.^2);
```

```
Vx = xxbar - xbar^2; Vy = yybar - ybar^2;
```

```
COVxy = xybar - xbar*ybar;
```

```
m = xybar/xxbar; r = COVxy/sqrt(Vx*Vy);
```

```
[x,i] = sort(x); y = y(i);
```

```
yfit = m*x; xfit = x;
```

```
%Compute error in gradient  $m$ 
```

```
n = length(x); s = sqrt( (1/(n-1))*sum( (y - yfit).^2 ) );
```

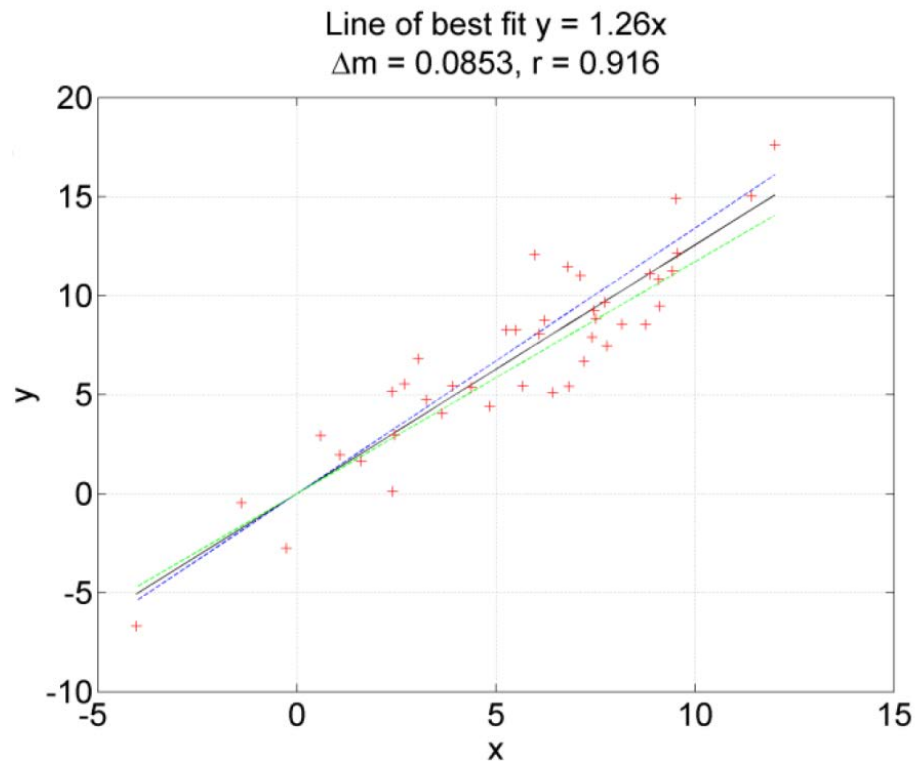
```
dm = s/sqrt(n*Vx);
```

```
%Determine envelope lines
```

```
yupper = (m+dm)*x; ylower = (m-dm)*x;
```

```
%Plot line of best fit
```

```
function plot_LOBF( x,y, yfit,xfit,r,m,dm,yupper,ylower,...  
    fsize, msize, xlabel_str, ylabel_str )  
figure('name','line of best fit','color',[1 1 1],...  
    'units','normalized','position',[0.05, 0.05, 0.9, 0.85]);  
plot( xfit, yfit, 'k-', xfit, yupper, 'b--', xfit, ylower, 'g--' );  
hold on; plot( x,y,'r+','markersize',msize); set(gca, 'fontsize',fsize );  
xlabel(xlabel_str); ylabel(ylabel_str); grid on;  
title( {'Line of best fit y = ( ',num2str(m,3), ' +/- ',num2str(dm,3),...  
    ' )x, r = ',num2str(r,4)] } );
```



Note this generic plot code is for a  $y = mx$  line of best fit