

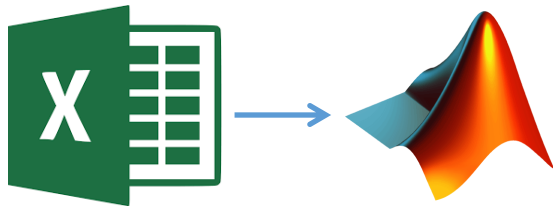
BPhO

Computational
Challenge

Seminar 02: An experimental data processing pipeline

Dr Andrew French.
December 2021.

Experimental data processing pipeline using Excel & MATLAB



Raw data in Excel

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

Perform analysis

- Averages
- Compute uncertainty
- Scaling
- Offset removal
- Linearization
- Line of best fit ...

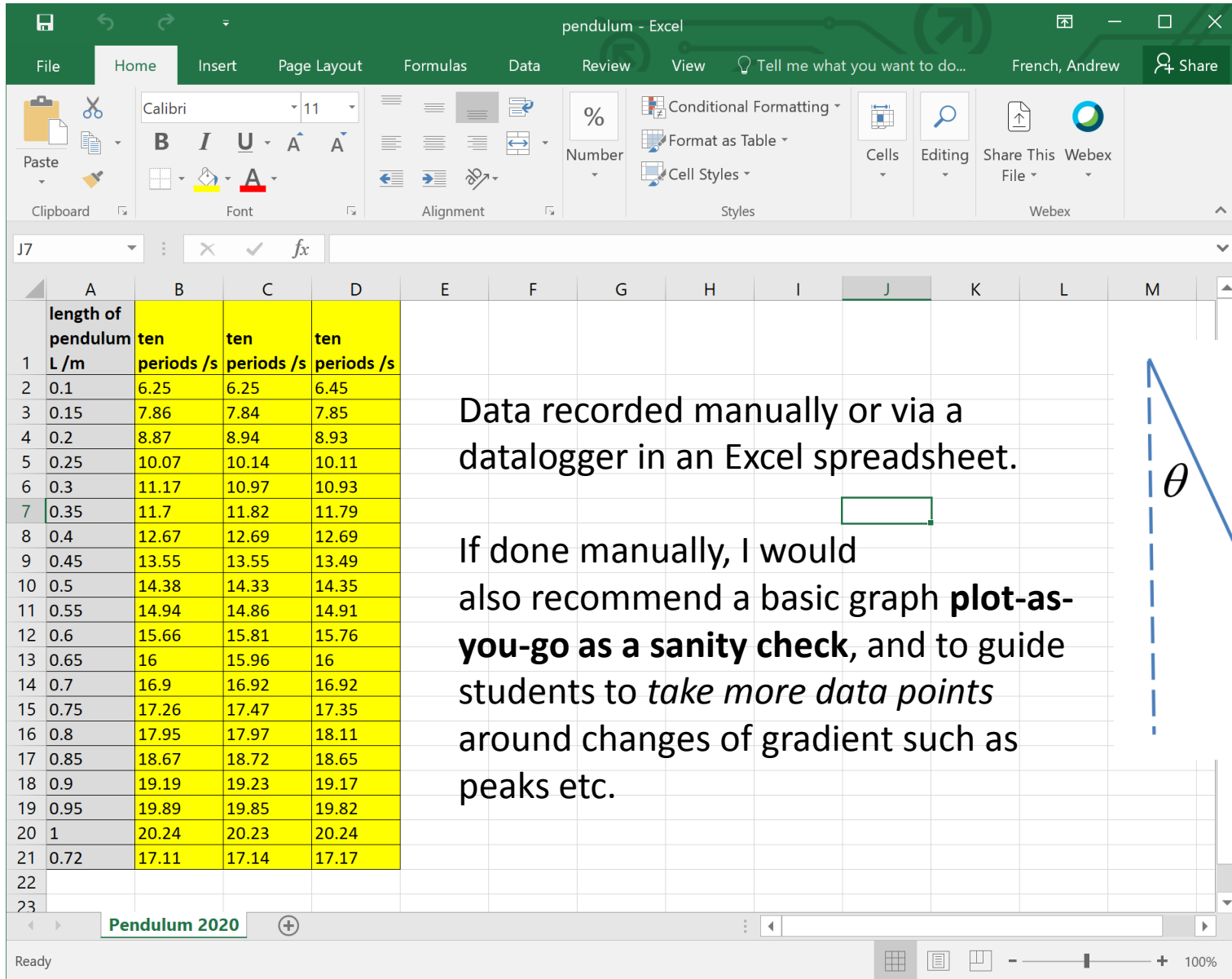
Plot data + error bars, *underlaid* with model curve

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

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

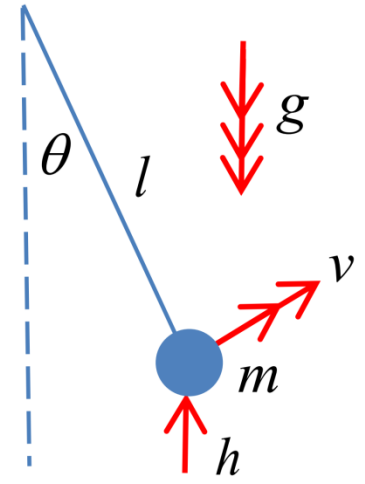
Example using pendulum data

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.



$$ml\ddot{\theta} = -mg \sin \theta \approx -mg\theta \quad \text{NEWTON II}$$

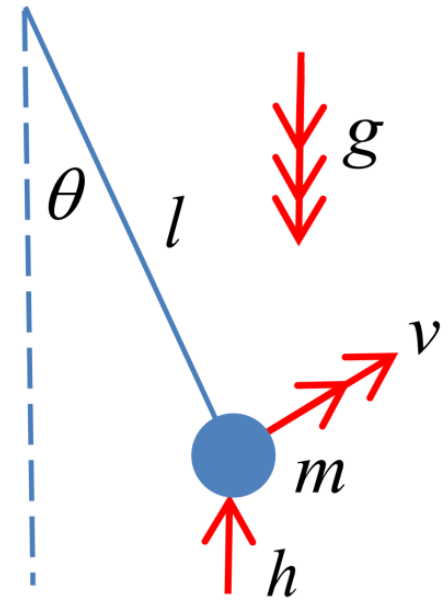
$$\therefore \ddot{\theta} = -\frac{g}{l}\theta$$

T is the pendulum period

$$\text{SHM: } \ddot{\theta} = -\left(\frac{2\pi}{T}\right)^2 \theta = -\omega^2 \theta$$

$$\theta(t) = \theta_0 \cos\left(\frac{2\pi t}{T}\right) = \theta_0 \cos \omega t$$

$$\therefore T = 2\pi \sqrt{\frac{l}{g}} \quad T^2 = 4\pi^2 \frac{l}{g} \quad g T^2 = \underbrace{4\pi^2 l}_y \quad \text{i.e. } y = gx$$



Simple Harmonic Motion (SHM) of a pendulum

* Ignore air resistance

* Small angle approximation i.e. $\theta \ll 1$ radian

length of pendulum L / m	ten periods /s	ten periods /s	ten periods /s	ten periods /s	period T /s	x = T ²	y = 4*π ² * L
0.1	6.25	6.25	6.45	6.32	0.63	0.40	3.95
0.15	7.86	7.84	7.85	7.85	0.79	0.62	5.92
0.2	8.87	8.94	8.93	8.91	0.89	0.79	7.90
0.25	10.07	10.14	10.11	10.11	1.01	1.02	9.87
0.3	11.17	10.97	10.93	11.02	1.10	1.22	11.84
0.35	11.7	11.82	11.79	11.77	1.18	1.39	13.82
0.4	12.67	12.69	12.69	12.68	1.27	1.61	15.79
0.45	13.55	13.55	13.49	13.53	1.35	1.83	17.77
0.5	14.38	14.33	14.35	14.35	1.44	2.06	19.74
0.55	14.94	14.86	14.91	14.90	1.49	2.22	21.71
0.6	15.66	15.81	15.76	15.74	1.57	2.48	23.69
0.65	16	15.96	16	15.99	1.60	2.56	25.66
0.7	16.9	16.92	16.92	16.91	1.69	2.86	27.63
0.75	17.26	17.47	17.35	17.36	1.74	3.01	29.61
0.8	17.95	17.97	18.11	18.01	1.80	3.24	31.58
0.85	18.67	18.72	18.65	18.68	1.87	3.49	33.56
0.9	19.19	19.23	19.17	19.20	1.92	3.69	35.53
0.95	19.89	19.85	19.82	19.85	1.99	3.94	37.50
1	20.24	20.23	20.24	20.24	2.02	4.10	39.48
0.72	17.11	17.14	17.17	17.14	1.71	2.94	28.42

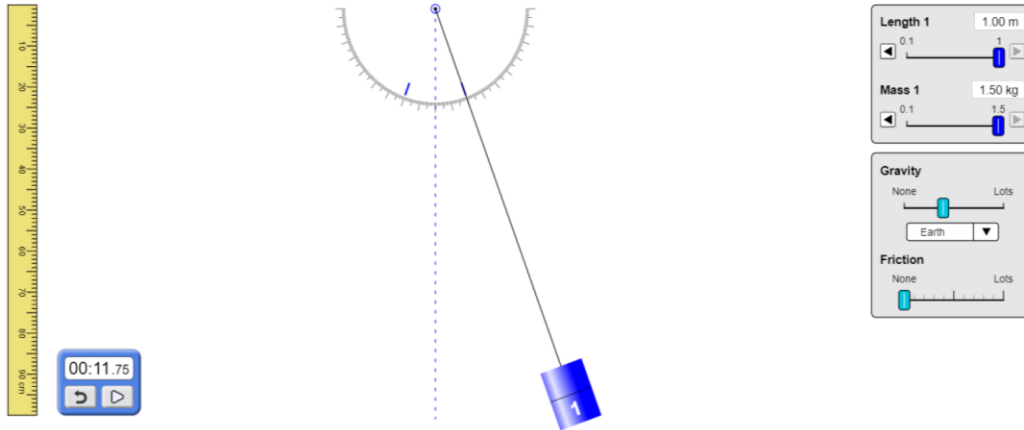
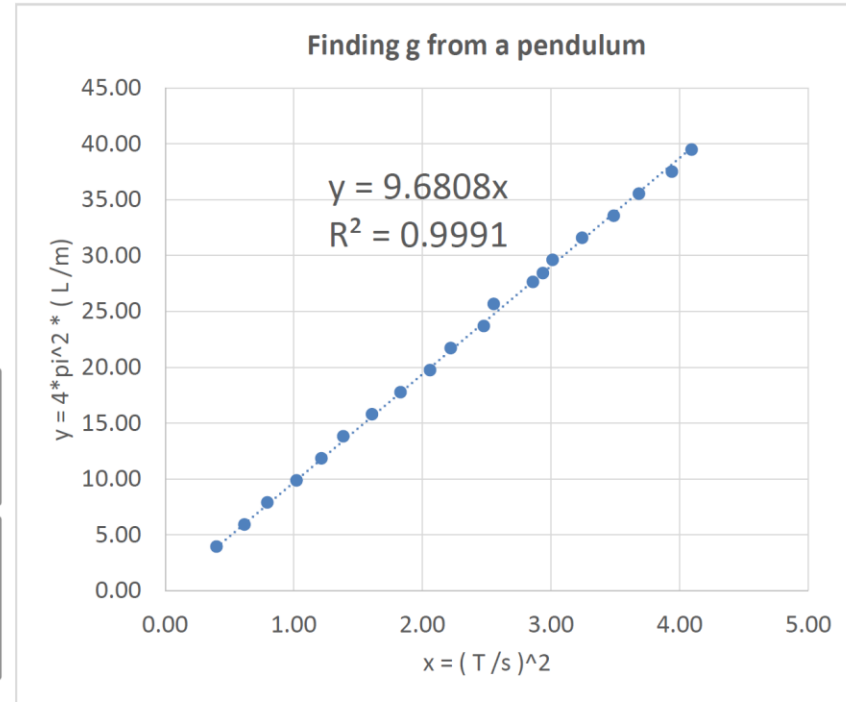
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$$\theta(t) = \theta_0 \cos\left(\frac{2\pi t}{T}\right) = \theta_0 \cos \omega t$$

$$\therefore T = 2\pi \sqrt{\frac{l}{g}} \quad T^2 = 4\pi^2 \frac{l}{g} \quad g T^2 = 4\pi^2 l \quad \text{i.e. } y = gx$$



Initial angle = 20 degrees

To complete, underlay (Period vs pendulum length) data with a model curve

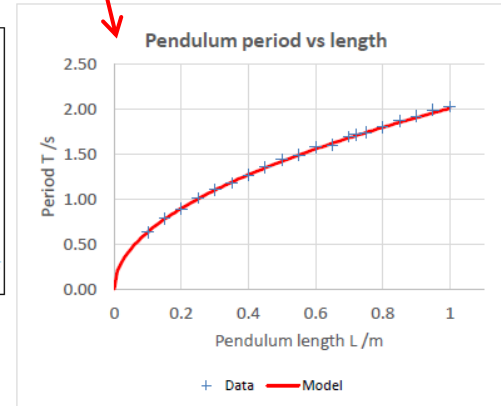
$$T = 2\pi \sqrt{\frac{l}{g}}$$

MEASURING g VIA A PENDULUM

03/06/2020

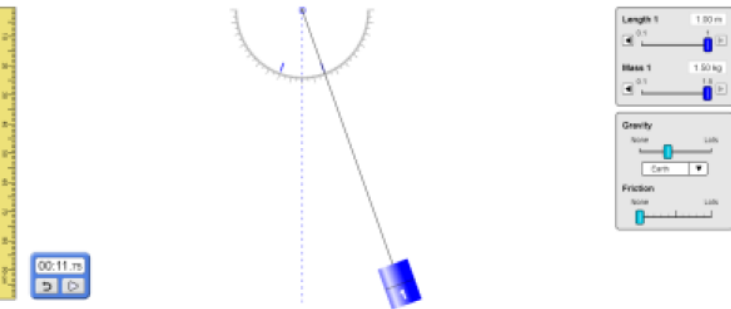
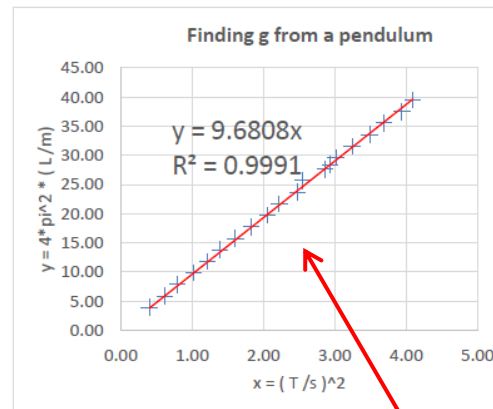
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 $\theta(t) = \theta_0 \cos\left(\frac{2\pi t}{T}\right) = \theta_0 \cos \omega t$
 $\therefore T = 2\pi \sqrt{\frac{l}{g}} \quad T^2 = 4\pi^2 \frac{l}{g} \quad g T^2 = 4\pi^2 l \quad \text{i.e. } y = gx$



Using $g = 9.81 \text{ N/kg}$
MODEL curve

L/m	T/s
0.000	0.000
0.010	0.201
0.020	0.284
0.030	0.347
0.040	0.401
0.050	0.449
0.060	0.491
0.070	0.531
0.080	0.567
0.090	0.602
0.100	0.634
0.110	0.665
0.120	0.695
0.130	0.723
0.140	0.751
0.150	0.777
0.160	0.802
0.170	0.827
0.180	0.851
0.190	0.874
0.200	0.897
0.210	0.919
0.220	0.941
0.230	0.962
0.240	0.983
0.250	1.003
0.260	1.023
0.270	1.042
0.280	1.062
0.290	1.080
0.300	1.099
0.310	1.117
0.320	1.135
0.330	1.152
0.340	1.170
0.350	1.187
0.360	1.204
0.370	1.220

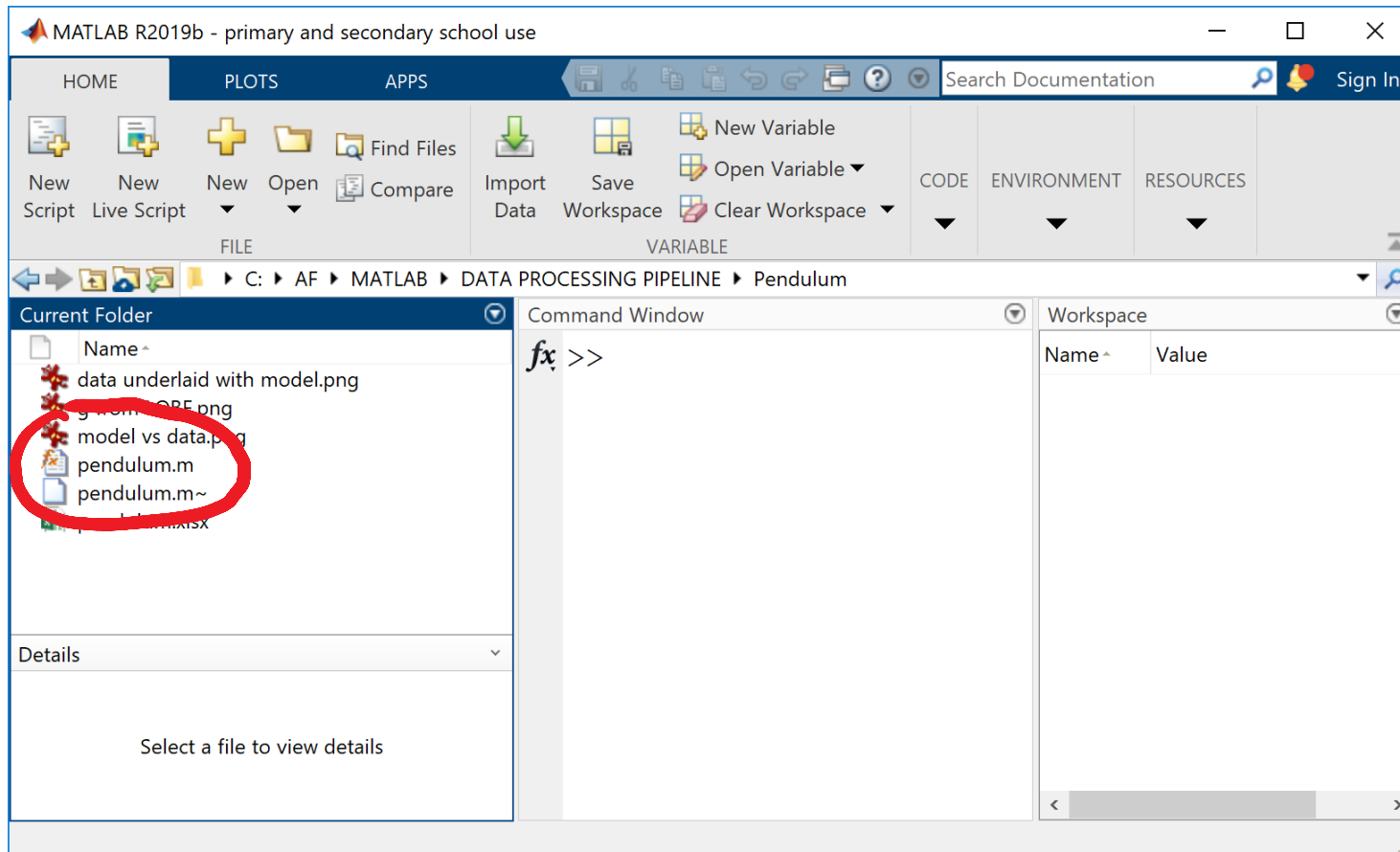


Initial angle = 20 degrees

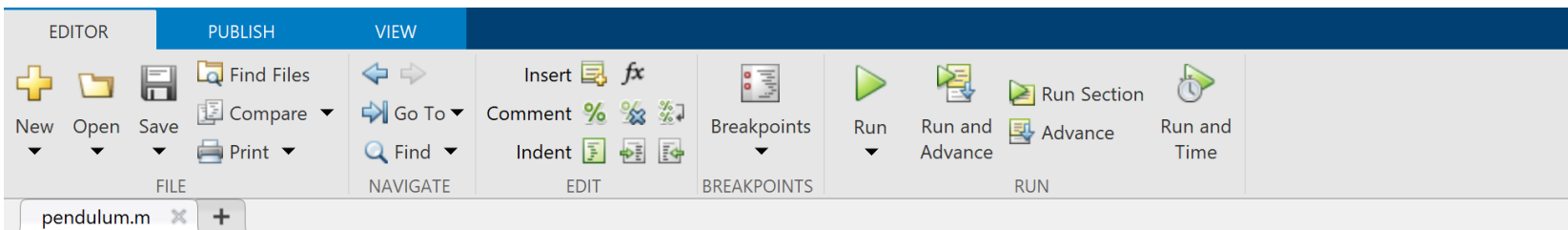
Linearization, line of best fit to assess model correlation, and determine g

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.



MATLAB
data
processing
pipeline



```

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;

```


	A	B	C	D	E	F	G	H	I	J	K	L	M
	length of pendulum	ten	ten	ten									
1	L/m	periods /s	periods /s	periods /s									
2	0.1	6.25	6.25	6.45									
3	0.15	7.86	7.84	7.85									
4	0.2	8.87	8.94	8.93									
5	0.25	10.07	10.14	10.11									
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12	0.6	15.66	15.81	15.76									
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14	0.7	16.9	16.92	16.92									
15	0.75	17.26	17.47	17.35									
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18	0.9	19.19	19.23	19.17									
19	0.95	19.89	19.85	19.82									
20	1	20.24	20.23	20.24									
21	0.72	17.11	17.14	17.17									
22													
23													

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

29

30

31

32

33

34

35

36

37

38

39

40

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

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

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

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

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

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

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

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

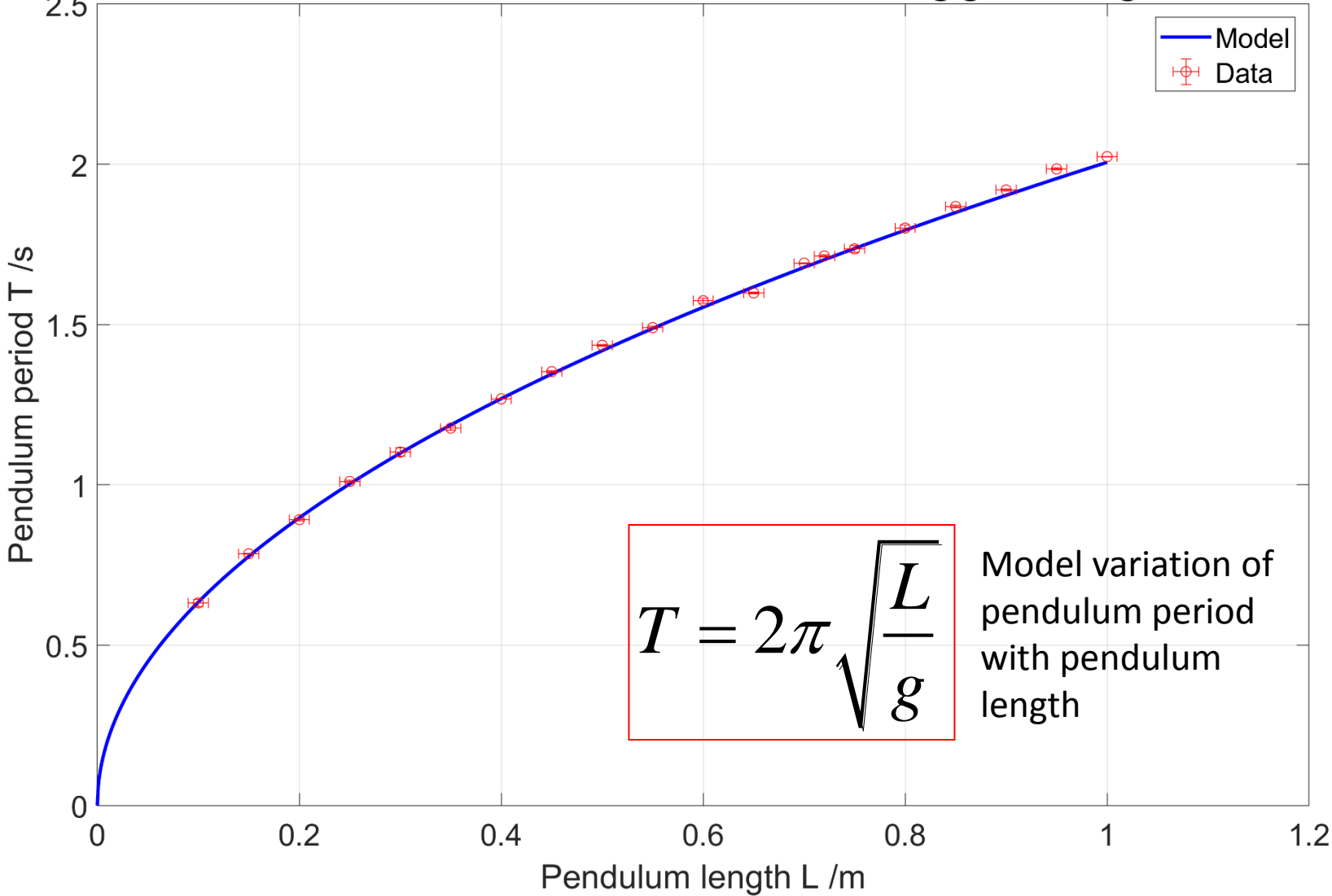
```

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 $g=9.81\text{N/kg}$



```

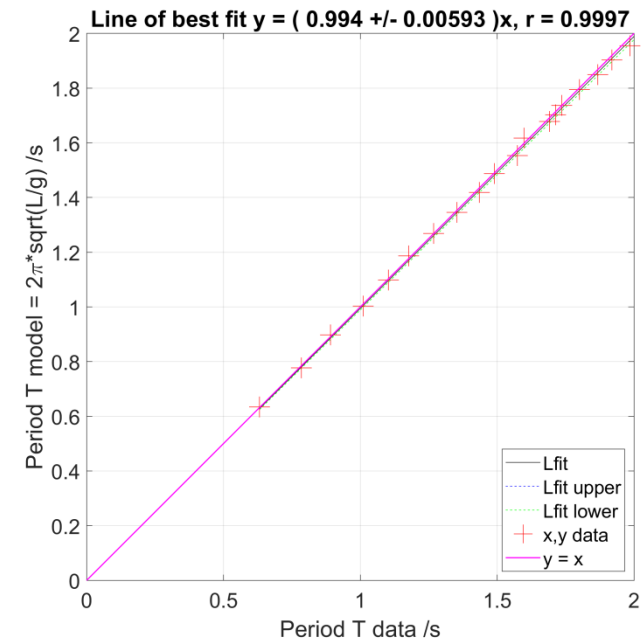
74 %% ANALYSIS: Determine line of best fit of the form y = m*x between T data and T model
75 % For 100% correlation, the gradient m = 1 and product-moment correlation coefficient r = 1.
76 y = Tmodel; x = T; [yfit,xfit,r,m,dm,yupper,ylower,s] = bestfit( x,y );
77
78 %Plot line of best fit
79 xlabel_str = '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

```

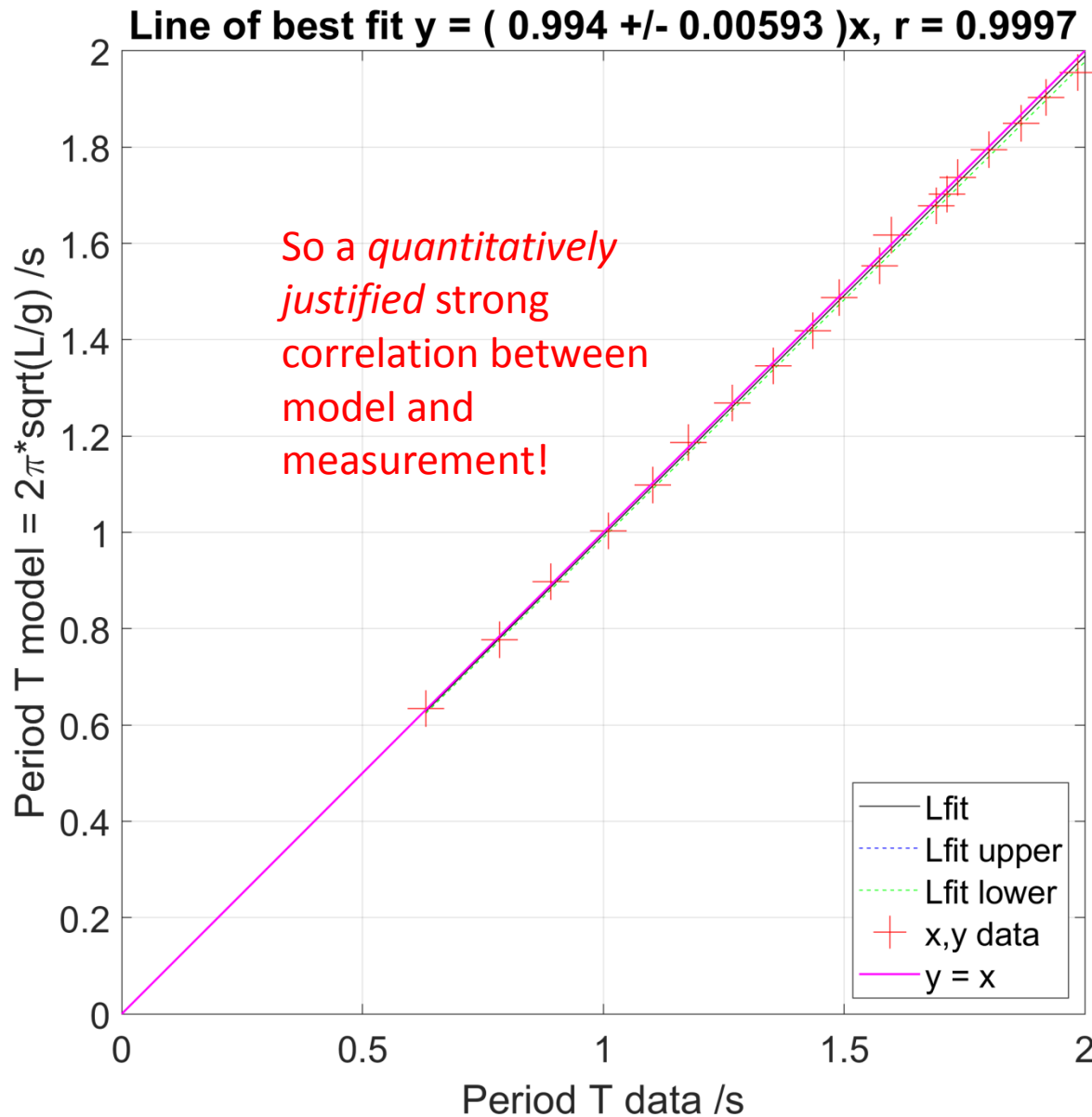
These are *sub-functions* which perform the line of best fit and associated plots. They should be generic, regardless of the dataset.

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

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

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

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

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)

```
%% 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');
```

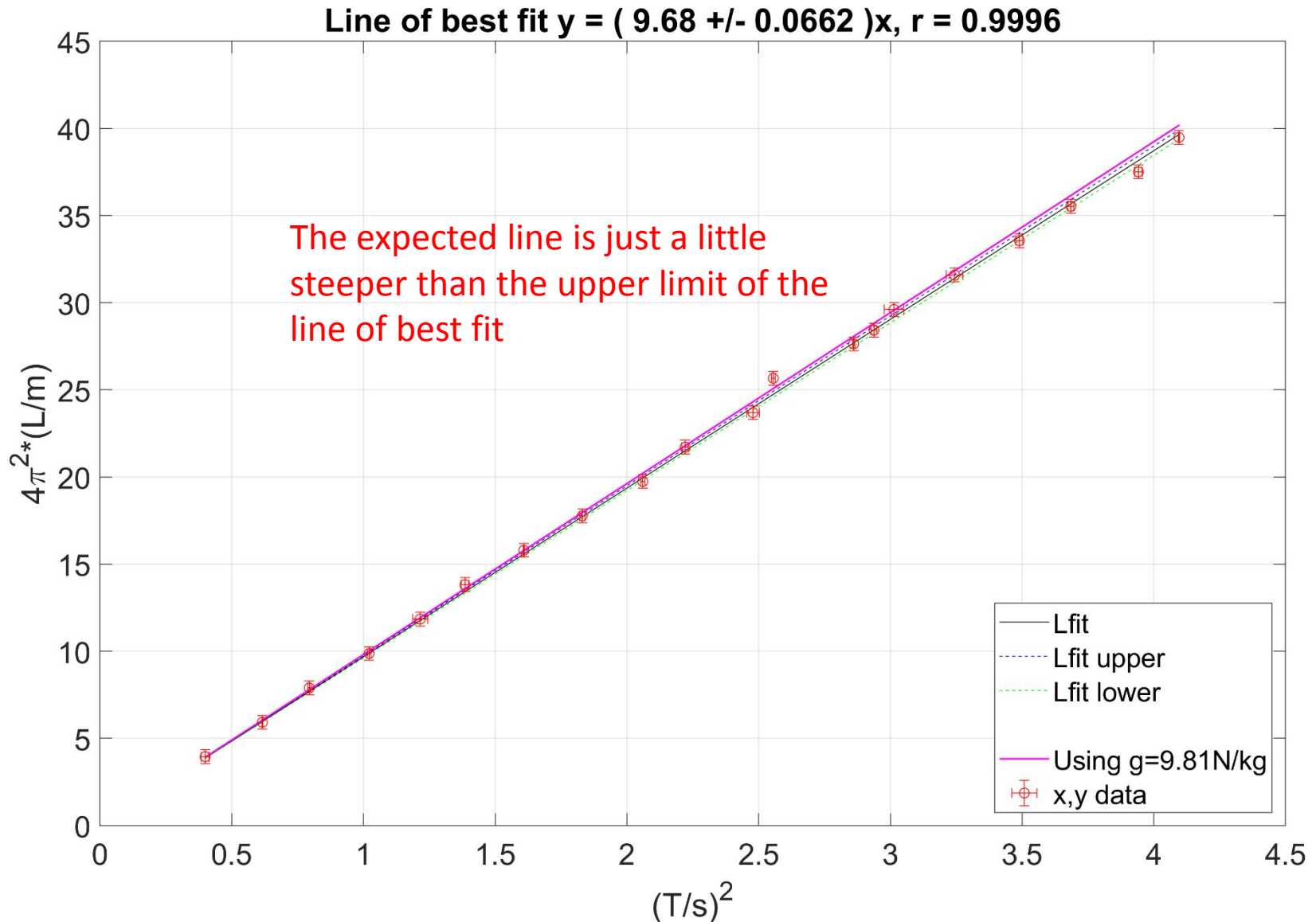
```
%Add a legend
```

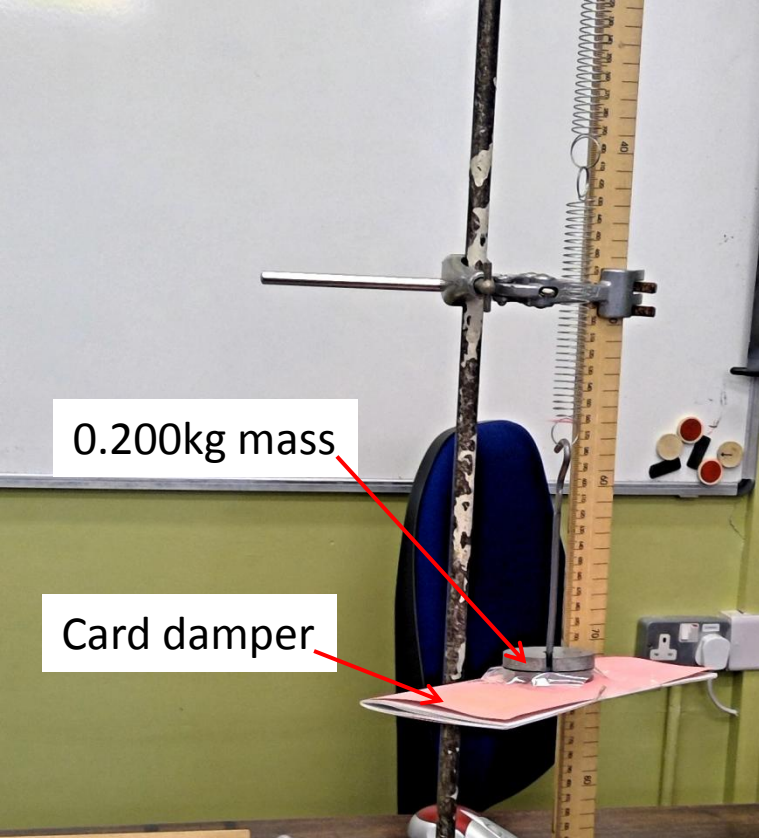
```
legend({'Lfit', 'Lfit upper','Lfit lower','',...}
```

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

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

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

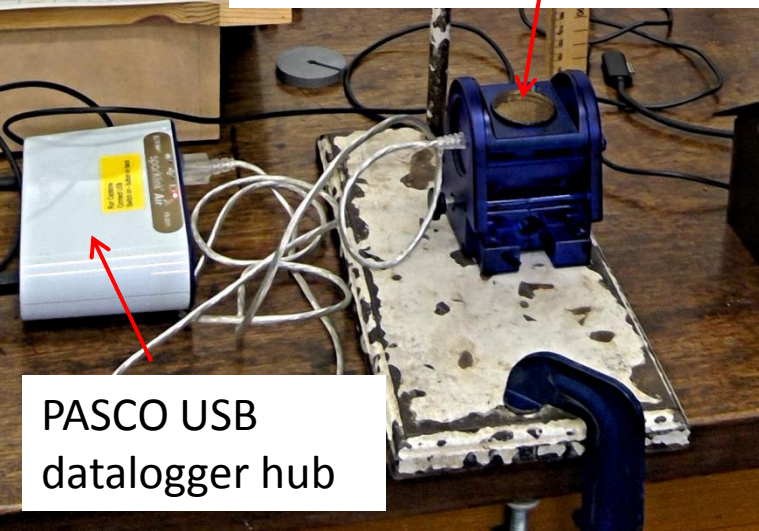




0.200kg mass

Card damper

Ultrasonic position sensor



PASCO USB
datalogger hub

Excel to MATLAB data processing pipeline example:

A **mass-spring system with damping**, with position recorded via an ultrasonic sensor and a datalogger.

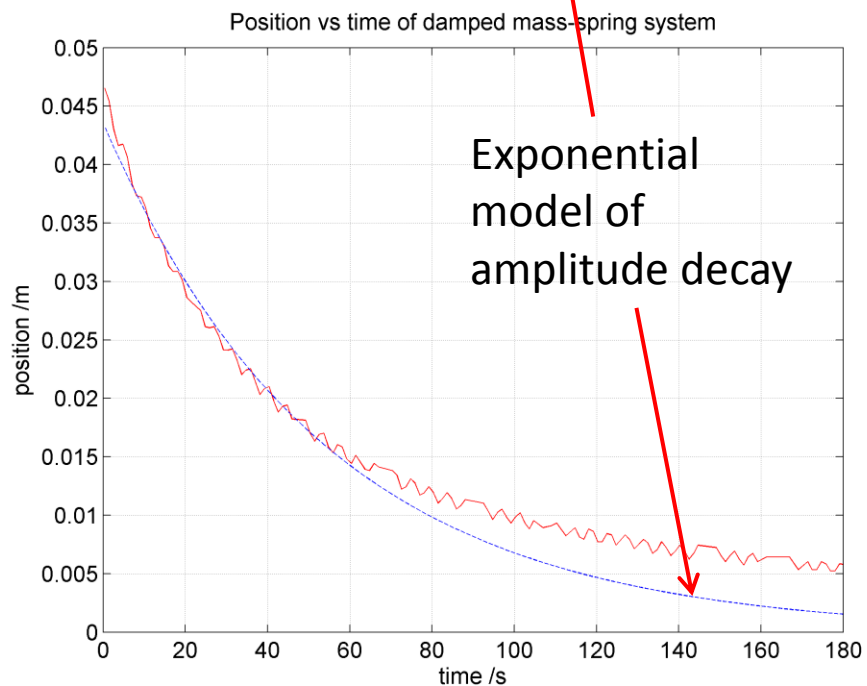
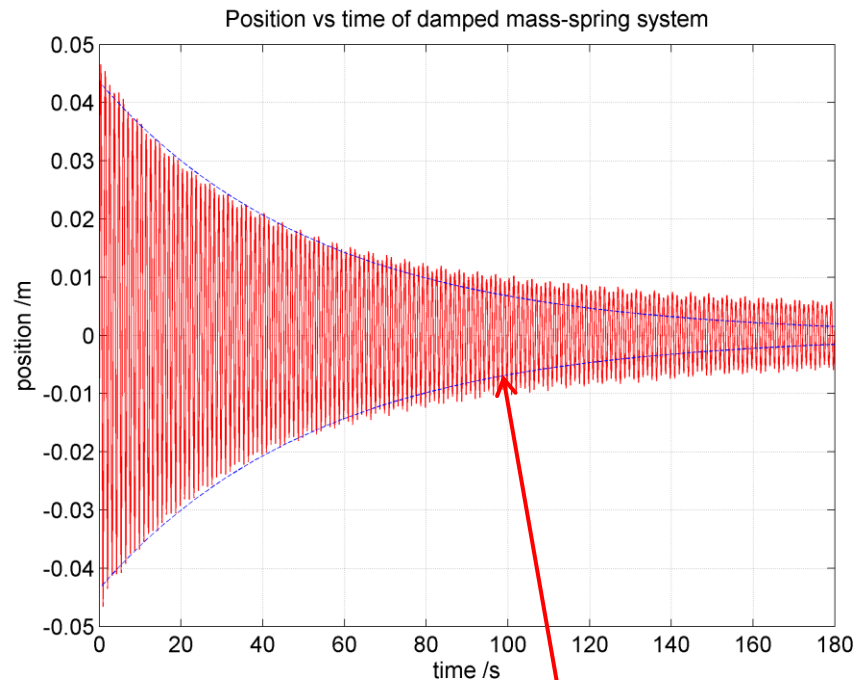
Mass-spring system with damping

0.200kg mass

Card damper

Ultrasonic position sensor

PASCO USB datalogger hub



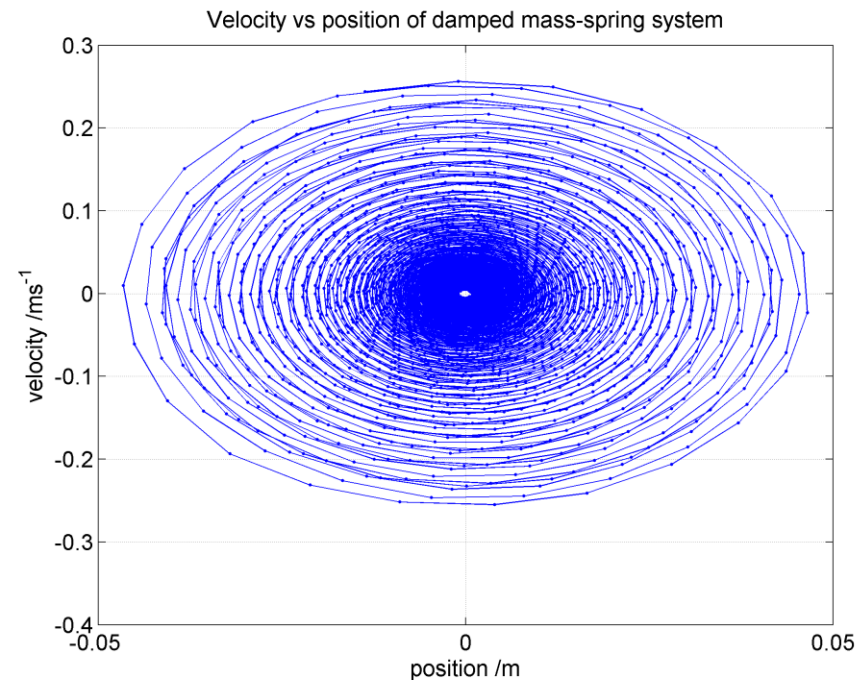
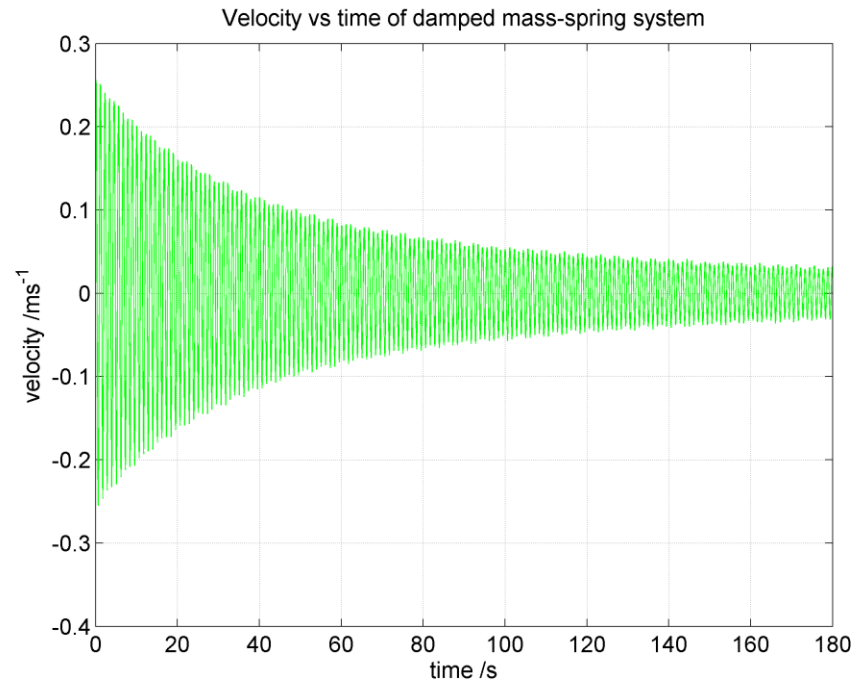
Mass-spring system with damping

0.200kg mass

Card damper

Ultrasonic position sensor

PASCO USB datalogger hub



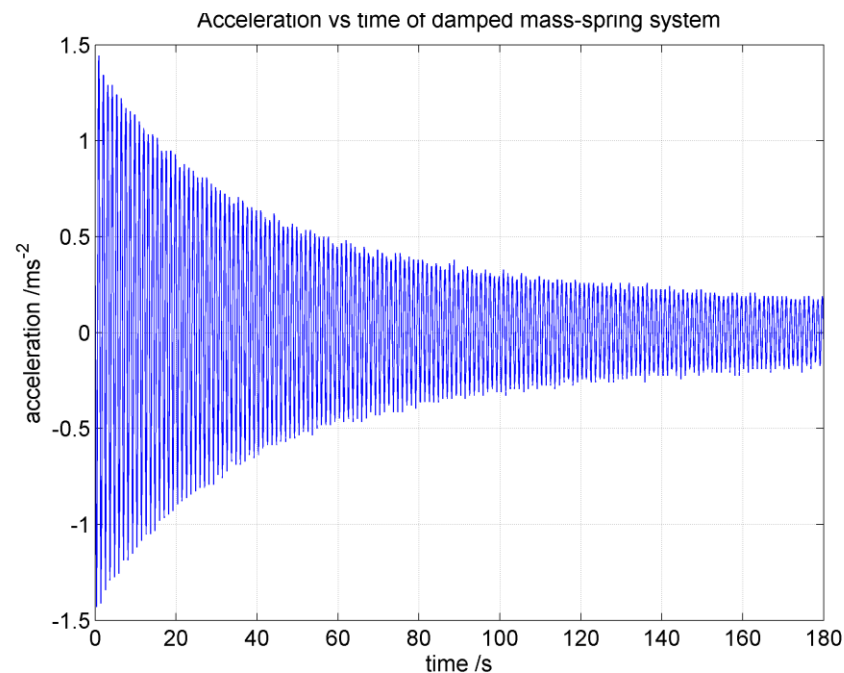
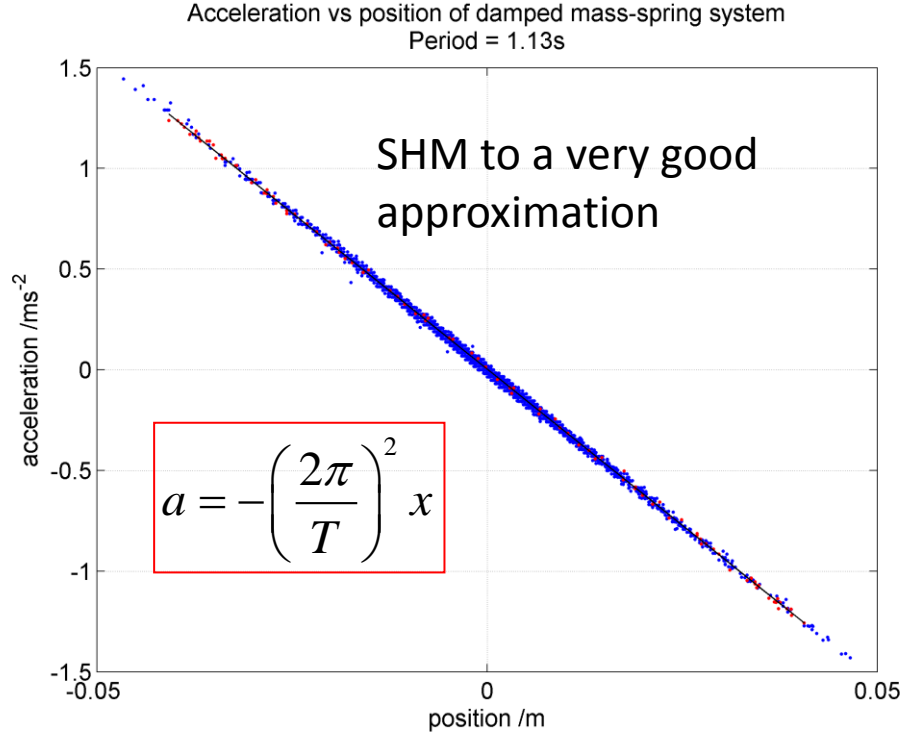
Mass-spring system with damping

0.200kg mass

Card damper

Ultrasonic position sensor

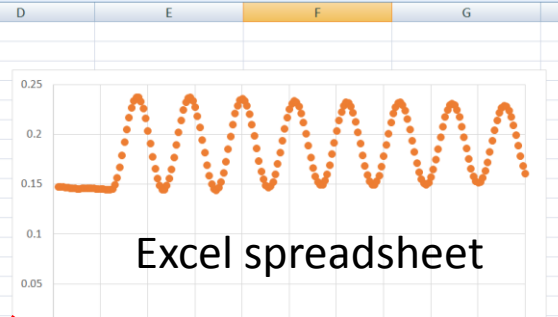
PASCO USB datalogger hub



Data processing pipeline for mass-spring system

Datalogger

1	Position (m)	Velocity (m/s)	Acceleration (m/s ²)	Time (s)
2	0.1479	1.48E-05	-0.0169	0.1
3	0.1479	-0.0017	-0.0516	0.15
4	0.1478	-0.0051	-0.0519	0.2
5	0.1474	-0.0069	-0.0176	0.25
6	0.1471	-0.0069	0.0343	0.3
7	0.1467	-0.0035	0.0522	0.35
8	0.1467	-0.0017	-0.0172	0.4
9	0.1465	-0.0052	-0.0181	0.45
10	0.1462	-0.0035	0.0861	0.5
11	0.1462	0.0034	0.0695	0.55
12	0.1465	0.0035	-0.0172	0.6
13	0.1465	0.0017	-0.0173	0.65
14	0.1467	0.0017	4.41E-04	0.7
15	0.1467	0.0018	-0.0512	0.75
16	0.1469	-0.0034	-0.1038	0.8
17	0.1464	-0.0086	-0.0351	0.85
18	0.146	-0.0069	0.0515	0.9
19	0.1457	-0.0035	0.0692	0.95

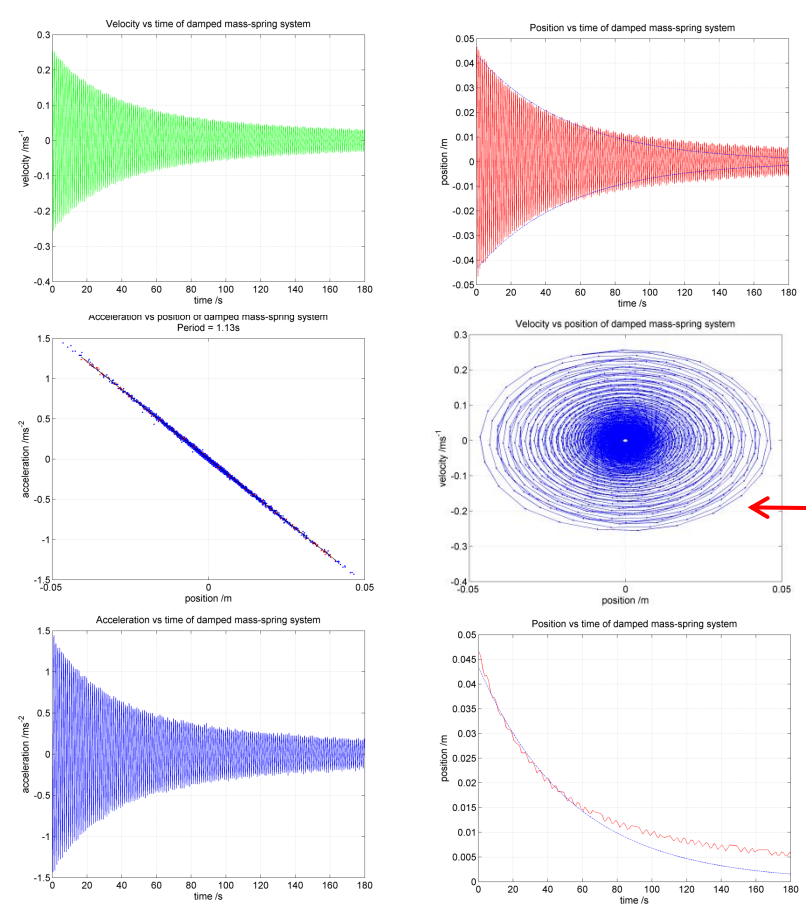


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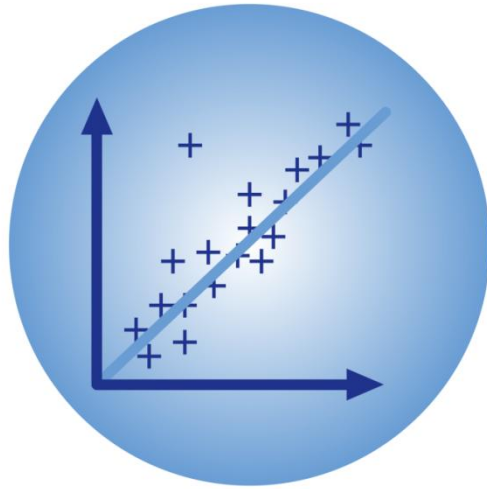
1 % Vertically mounted three-springs-in-series are attached to a 0.2kg mass
2 % and sellotaped to a card diary to increase air resistance.
3 % Position of the mass is determined via a PASCO ultrasonic transducer,
4 % and data is logged at 20Hz using the Capstone system. Capstone
5 % automatically determines velocity and acceleration via some form of
6 % numerical differentiation.
7  function damped_mass_on_spring_analysis
8 fsize = 14; %FontSize for graphs
9 tstart = 0; tend = 60*3; %Time interval
10
11 %Load data
12 [num,txt,row] = xlsread('Damped mass on spring.xlsx');
13
14 %Ignore first few data points (oscillation hasn't started yet)
15 num = num(50:end,:);
16
17 %Determine position, velocity, acceleration, time in SI units
18 t = num(:,4); t = t - t(1);
19 x = num(:,1); v = num(:,2); a = num(:,3);
20
21 %Determine equilibrium height from the average position, and shift x
22 %such that equilibrium displacement is zero.
23 x = x - mean(x);
24
25 %Restricted data set (i.e. where effect of damping can be ignored)
26 aa = a(100:200); tt = t(100:200); xx = x(100:200);
27
28 %Determine line of best fit between a and x to determine period
29 [m,c,yfit,r] = lfit(xx,aa); T = 2*pi/sqrt(-m);
30
31 %Find peaks and plot these to determine exponential envelope
32 [tp,xp] = peakfindergeneral(t,x); N = length(xp);
33
34 %Linearize exponential envelope. Choose which set of data you want to use
35 %for this computation. NOTE THE ENVELOPE IS NOT EXPONENTIAL IN THE SAME WAY
36 %DURING THE WHOLE DECAY, SINCE THE DRAG FORCE WON'T BE LINEAR WITH SPEED.
37 i = find(xp>0); i = i(1:50);
38 Y = log(xp(i)); X = tp(i);
39
40 %Determine exponential envelope in the form x = (+/-) A*exp(-k*t)
41 [m,c,YFIT,r] = lfit(X,Y);

```

MATLAB



Auto-generated graphs and analysis



BPhO

Computational Challenge

- Suggested homework
- Q&A