## **Physics 1 Lab**

# Density measurement using Archimedes' method and reaction time measurement



Physics Department

Parsa Hatami 400100962 Alireza Haghshenas 400101067

Lab 2

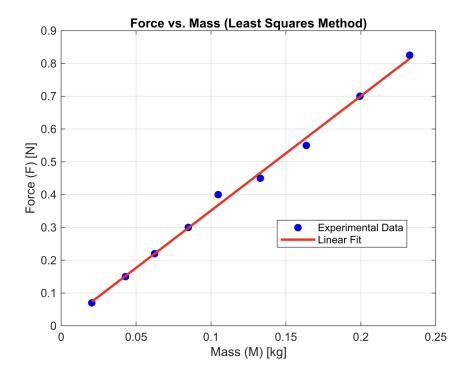
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## Density measurement using Archimedes' method

## **Question 1**

Plot the graph of M versus F, determine the **slope** and **y-intercept** using the least squares method.

```
T1 = [0.2 \ 0.42 \ 0.61 \ 0.83 \ 1.025 \ 1.300 \ 1.600 \ 1.950 \ 2.275 ];
 T2 = [0.13 \ 0.27 \ 0.39 \ 0.53 \ 0.625 \ 0.850 \ 1.050 \ 1.250 \ 1.450];
 \% Calculate Force F as the difference between T1 and T2
 M = T1/9.78;
 F = T1 - T2;
 % Perform linear regression (least squares method) for F vs. M
 p = polyfit(M, F, 1); % p(1) is the slope, p(2) is the intercept
 \% Generate fitted values based on the linear model
 F_fit = polyval(p, M);
13
 figure;
 plot(M, F, 'bo', 'MarkerFaceColor', 'b'); % Data points (blue circles)
 plot(M, F_fit, 'r-', 'LineWidth', 2); % Fitted line (red)
 xlabel('Mass (M) [kg]');
 ylabel('Force (F) [N]');
 title('Force vs. Mass (Least Squares Method)');
 legend('Experimental Data', 'Linear Fit', 'Location', 'best');
 grid on;
 fprintf('Linear Fit Equation: F = \%.2f*M + \%.2f*n', p(1), p(2));
 y_fit = polyval(p, M);
 SS_{tot} = sum((F - mean(F)).^2);
 SS_res = sum((F - y_fit).^2);
 R2 = 1 - (SS_{res} / SS_{tot});
 fprintf('R-squared: %.4f\n', R2);
```



#### **Linear Fit Equation:**

F = 3.49M + 0.00

**R-squared:** 0.9961

## **Question 2**

Calculate the **density slope** of the metal.

```
rho_water = 1003.5; % Density of water in kg/m^3
g = 9.81; % Gravitational acceleration in m/s^2

T1 = [0.2 0.42 0.61 0.83 1.025 1.300 1.600 1.950 2.275 ];
T2 = [0.13 0.27 0.39 0.53 0.625 0.850 1.050 1.250 1.450];

M = T1/9.78;

F = T1 - T2;

% Perform linear regression (least squares method) for F vs. M
p = polyfit(M, F, 1); % p(1) is the slope, p(2) is the intercept

slope = p(1);

rho_metal = (rho_water * g) / slope;

fprintf('The slope of the (F - M) line: %.4f\n', slope);
```

The slope of the (F - M) line: 3.4934

Calculated metal density: 2817.9446 kg/m<sup>3</sup>

## **Question 3**

Explain the significance of the **regression coefficient** and interpret how well the graph fits.

#### Answer

The **regression coefficient** (slope) of the line represents the rate of change of the dependent variable (force F) with respect to the independent variable (mass M) in your experiment.

Mathematically, the regression equation is:

$$F = \operatorname{slope} \times M + \operatorname{intercept}$$

The slope tells you how much F increases (or decreases) when M increases by one unit.

#### Interpretation:

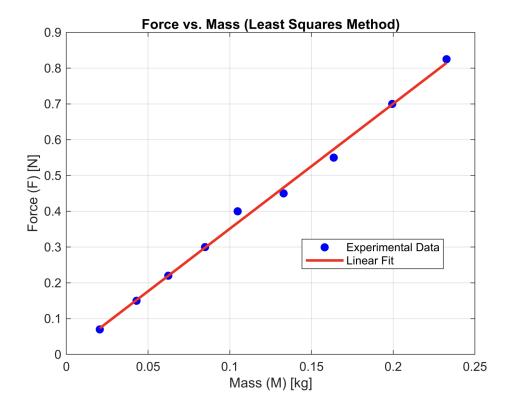
- A **positive slope** indicates that as the mass *M* increases, the force *F* also increases. This suggests a *directly proportional* relationship between the two variables, possibly due to the material or experimental setup.
- A **negative slope** would suggest an *inverse relationship*, meaning that as mass increases, the force decreases.
- The **magnitude** of the slope tells you the strength of this relationship. A larger value means that even small changes in mass lead to larger changes in force.

## **E**xample:

If the regression coefficient (slope) of your model is 0.5, then for every 1 kg increase in mass M, the force F would increase by 0.5 N. The units of the slope are important because they represent the *rate* of change (in this case, N/kg, Newtons per kilogram).

This provides a clear and direct understanding of how mass influences force in your experiment, based on the **linear model**.

```
12
 figure;
 plot(M, F, 'bo', 'MarkerFaceColor', 'b'); % Data points (blue circles)
13
 hold on;
 plot(M, F_fit, 'r-', 'LineWidth', 2); % Fitted line (red)
 xlabel('Mass (M) [kg]');
 ylabel('Force (F) [N]');
 title('Force vs. Mass (Least Squares Method)');
 legend('Experimental Data', 'Linear Fit', 'Location', 'best');
 grid on;
 fprintf('Linear Fit Equation: F = \%.2f*M + \%.2f\n', slope, intercept);
 fprintf('The regression coefficient (slope) is %.2f.\n', slope);
25
 if slope > 0
      fprintf('This indicates a positive relationship: as mass (M) increases,
         force (F) increases.\n');
 else
28
      fprintf('This indicates a negative relationship: as mass (M) increases,
29
         force (F) decreases.\n');
 end
```



Linear Fit Equation: F = 3.49M + 0.00 Regression Coefficient (Slope): 3.49

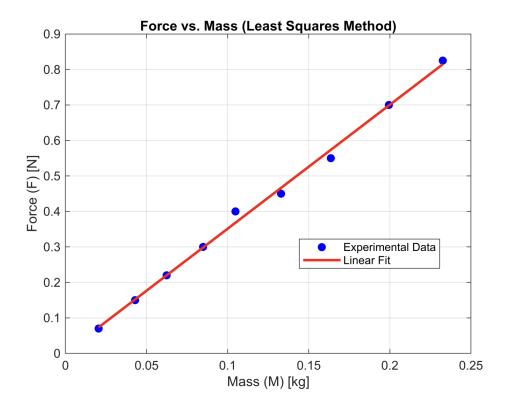
This indicates a positive relationship: as mass (M) increases, force (F) increases.

## **Question 4**

Using the slope and its uncertainty, determine the error in measuring the metal density.

error in slope = 
$$\sqrt{\frac{\sum (y - y_{\text{fit}})^2}{n - 2}} \cdot \frac{1}{\sqrt{\sum (x - \overline{x})^2}}$$

```
\% Perform linear regression (least squares method) for F vs. M
 [p, S] = polyfit(M, F, 1); % p(1) is the slope, p(2) is the intercept, S
     is the structure with the error info
 % Extract slope and intercept from the linear regression
 slope = p(1);
 intercept = p(2);
 \% Get the standard error of the slope and intercept from the structure S
 slope_error = sqrt(S.normr^2 / (length(M) - 2)) * sqrt(1 / sum((M - mean(M)))
     .^2));
 % Generate fitted values based on the linear model
 F_fit = polyval(p, M);
 figure;
 plot(M, F, 'bo', 'MarkerFaceColor', 'b'); % Data points (blue circles)
 plot(M, F_fit, 'r-', 'LineWidth', 2); % Fitted line (red)
 xlabel('Mass (M) [kg]');
 ylabel('Force (F) [N]');
 title('Force vs. Mass (Least Squares Method)');
 legend('Experimental Data', 'Linear Fit', 'Location', 'best');
 grid on;
22
23
 % Display the linear regression parameters (slope, intercept, and their
 fprintf('Linear Fit Equation: F = %.2f*M + %.2f\n', slope, intercept);
 fprintf('Slope error: %.4f\n', slope_error);
```



**Linear Fit Equation:** F = 3.49M + 0.00

Slope Error: 0.0822

## **Question 5**

Compare the **intercept error** obtained from the graph with the calculated value.

```
[p, S] = polyfit(M, F, 1);  % p(1) is the slope, p(2) is the intercept, S
    is the structure with the error info

% Extract slope and intercept from the linear regression
slope = p(1);
intercept = p(2);

% Get the standard error of the slope and intercept from the structure S
slope_error = sqrt(S.normr^2 / (length(M) - 2)) * sqrt(1 / sum((M - mean(M)))
    .^2));

% Suppose you have an expected error value, replace this with the actual
value:
expected_error =0.02;  % Set expected error here

fprintf('Linear Fit Equation: F = %.2f*M + %.2f\n', slope, intercept);
```

```
fprintf('Slope error from regression: %.4f\n', slope_error);

fprintf('Slope error from regression: %.4f\n', slope_error);

error_difference = abs(slope_error - expected_error);

fprintf('Difference between calculated error and expected error: %.4f\n', error_difference);

if error_difference < 0.1
    fprintf('The error is within the acceptable range.\n');

else
    fprintf('The error is outside the acceptable range.\n');
end</pre>
```

**Linear Fit Equation:** F = 3.49M + 0.00

**Slope Error from Regression:** 0.0822

**Difference Between Calculated and Expected Error:** 0.0622

**Conclusion:** The error is within the acceptable range.

## Reaction time measurement of a person

## **Question 1, Question 2**

Plot the distribution for the given data based on tables **H2** and **H3**.

Calculate the **standard deviation** and **mean** for tables **H2** and **H3**, and explain their significance. (You may also use curve-fitting software such as *TableCurve*, *Origin*, *SigmaPlot*.)

#### Answer for Question 1,2

```
time_H2 = [199 185 465 167 172 245 387 401 10 132 180 162 204 15 127 4 183
     162 220 119]; % Time delays for H2 (Alireza.H)
 time_H3 = [165 215 156 186 293 267 189 192 69 75 181 410 6 179 186 175 25
     350 184 159]; % Time delays for H3 (Parsa.H)
 table_H2 = table((1:length(time_H2))', time_H2', 'VariableNames', {'try', '
     Time_H2'});
 disp('Table for H2:');
 disp(table_H2);
 table_H3 = table((1:length(time_H3))', time_H3', 'VariableNames', {'try', '
     Time_H3'});
 disp('Table for H3:');
 disp(table_H3);
 mean_H2 = mean(time_H2);
 std_H2 = std(time_H2);
 fprintf('For H2 Group:\n');
 fprintf('Mean response time: %.2f ms\n', mean_H2);
 fprintf('Standard deviation: %.2f ms\n', std_H2);
 % Basic Statistical Analysis for H3 Group
 mean_H3 = mean(time_H3);
 std_H3 = std(time_H3);
 fprintf('For H3 Group:\n');
 fprintf('Mean response time: %.2f ms\n', mean_H3);
 fprintf('Standard deviation: %.2f ms\n', std_H3);
 figure;
31
 subplot(2,1,1);
 plot(1:length(time_H2), time_H2, 'bo-', 'LineWidth', 2, 'MarkerSize', 8);
 title('Response Time for H2 Group');
 xlabel('try');
 ylabel('Time Delay (ms)');
 grid on;
 subplot(2,1,2);
```

```
40 plot(1:length(time_H3), time_H3, 'ro-', 'LineWidth', 2, 'MarkerSize', 8);
 title('Response Time for H3 Group');
 xlabel('try');
 ylabel('Time Delay (ms)');
 grid on;
46
 figure;
47
 subplot(2,1,1);
 histogram(time_H2, 'BinWidth', 10, 'FaceColor', 'b', 'EdgeColor', 'k', '
     Normalization', 'pdf');
 hold on;
 % Fit a normal distribution to the data
pd_H2 = fitdist(time_H2', 'Normal');
s2 x_H2 = linspace(min(time_H2), max(time_H2), 100);
y_{H2} = pdf(pd_{H2}, x_{H2});
 plot(x_H2, y_H2, 'r-', 'LineWidth', 2); % Plot fitted PDF
 title('Histogram and Normal Fit for H2');
 xlabel('Time Delay (ms)');
 ylabel('Probability Density');
 grid on;
 hold off;
60
 subplot (2,1,2);
62
 histogram(time_H3, 'BinWidth', 10, 'FaceColor', 'r', 'EdgeColor', 'k', '
     Normalization', 'pdf');
 hold on;
64
65
 pd_H3 = fitdist(time_H3', 'Normal');
67 x_H3 = linspace(min(time_H3), max(time_H3), 100);
y_H3 = pdf(pd_H3, x_H3);
 plot(x_H3, y_H3, 'b-', 'LineWidth', 2); % Plot fitted PDF
 title('Histogram and Normal Fit for H3');
 xlabel('Time Delay (ms)');
 ylabel('Probability Density');
 grid on;
 hold off;
```

try	Time_H2
1	199
2	185
3	465
4	167
5	172
6	245
7	387
8	401
9	10
10	132
11	180
12	162
13	204
14	15
15	127
16	4
17	183
18	162
19	220
20	119

try	Time_H3
1	165
2	215
3	156
4	186
5	293
6	267
7	189
8	192
9	69
10	75
11	181
12	410
13	6
14	179
15	186
16	175
17	25
18	350
19	184
20	159

Table 1: Response Time Data for H2

Table 2: Response Time Data for H3

#### For H2 Group:

Mean response time: 186.95 ms Standard deviation: 120.43 ms

#### For H3 Group:

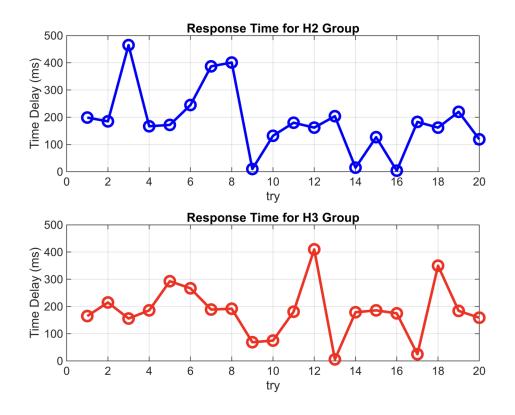
Mean response time: 183.10 ms Standard deviation: 97.98 ms

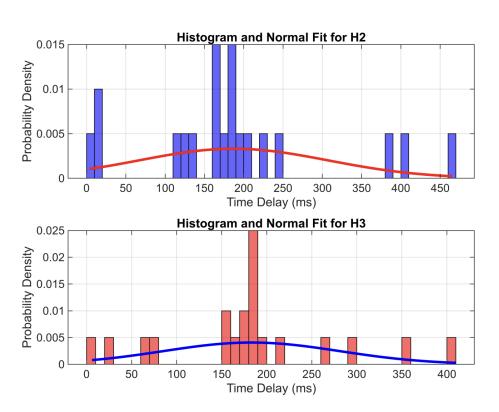
**Understanding the Mean:** The mean response time represents the *average* reaction time recorded in each experiment. Since the mean values for H2 (186.95 ms) and H3 (183.10 ms) are **very close**, we can say that, on average, both groups responded in nearly the same amount of time.

**Understanding Standard Deviation:** Standard deviation tells us how much the individual response times vary from the mean.

- The **H2 group** has a standard deviation of **120.43 ms**, meaning there is a **wider spread** in response times—some tries reacted much faster or slower.
- The **H3 group** has a standard deviation of **97.98 ms**, meaning the responses were **more consistent** and closer to the average.

In simpler terms, the H2 group had **more variation**, whereas the H3 group was **more stable** in their reaction times.





## **Question 3**

Are the **statistical behaviors** of the data from the two experiments different?

#### Answer

The two experiments show **different statistical behaviors** but both obey from normal distributions with near mean but some different standard deviations.

#### How much do the reaction times vary?

The **H2 group** had a **higher standard deviation**, meaning that the response times varied significantly. Some tries reacted very quickly, while others took much longer. The **H3 group**, on the other hand, had a lower standard deviation, meaning that **most responses were clustered closer to the mean**. This suggests that the H3 group was more consistent in their reaction times.

#### **Shape of the Data Distribution**

Looking at the histograms:

- The **H2 group's histogram** is more **spread out**, with reaction times ranging from very low (4 ms) to very high (465 ms).
- The **H3 group's histogram** is more **concentrated**, meaning that most participants responded within a smaller time range.

#### Are there any extreme values (outliers)?

In the H2 group, there were **several extreme values**, meaning that some individuals took much longer or much shorter time to respond. In contrast, the H3 group had **fewer extreme values**, meaning the reaction times were more stable.

Although the **average response time** for both groups was nearly the same, the way people responded was different:

- **H2 group:** More variation in response times, more extreme values, and a less predictable pattern.
- **H3 group:** More consistent responses, with most participants reacting within a similar time-frame.

In summary, while both groups had similar average response times, **H2 was more erratic, while H3** followed a more structured and predictable pattern.

## **Tables for Experiment 2**

## Table 1 - Density Measurement Data

M	1	2	3	4	5	6	7	8	10
$T_1$	0.200	0.420	0.610	0.830	1.025	1.300	1.600	1.950	2.275
$T_2$	0.130	0.270	0.390	0.530	0.625	0.850	1.050	1.250	1.450
$B=T_1-T_2$	0.070	0.150	0.220	0.300	0.400	0.450	0.550	0.700	0.825

Table 3: Density Measurement using Archimedes' Principle

## Table 2 - Reaction Time Measurement (First Subject)

Try	1	2	3	4	5	6	7	8	9	10
Time (ms)	199	185	465	167	172	245	387	401	10	132
Time (ms)	180	162	204	15	127	4	183	162	220	119

Table 4: Reaction Time Data for First Subject

## Table 3 - Reaction Time Measurement (Second Subject)

Try	1	2	3	4	5	6	7	8	9	10
Time (ms)	165	215	156	182	293	267	189	192	69	75
Time (ms)	181	410	6	179	186	175	25	350	184	159

Table 5: Reaction Time Data for Second Subject