

Islamic University of Technology (IUT)
Organization of Islamic Cooperation (OIC)
Department of Electrical and Electronic Engineering (EEE)

COURSE NO : **EEE 4416**
LAB NO : **08 (Part B)**
TOPIC : **DATA HANDLING USING MATLAB**

DATA HANDLING USING MATLAB

- ⇒ In this course, we are mostly interested in structured (tabular) data. In later courses, you will learn how to work with other unstructured data.
- ⇒ In this lab, you will learn the following things.

- How to import data into MATLAB?
- How to export data from MATLAB.
- Data exploration
- Data cleaning
- Data manipulation
- Data analysis
- Data visualization (will be covered in lab 09)

Data Importing

There are two different ways that you can use to import data into MATLAB.

- I. You can use the ‘Import Data’ section from the ‘Home’ tab to directly import data.
- II. You can use different built-in commands to load different types of data.

MATLAB supports importing data from various formats:

- **Text files** (.txt, .csv) → readtable, readmatrix, importdata, textscan, fileread
- **Excel files** (.xlsx) → readtable, xlsread
- **MAT-files** (.mat) → load
- **Images** → imread
- **Audio** → audioread
- **Web data, JSON, XML** → webread, jsondecode, xmlread

Use the above commands to read different data. Make sure the data files are stored in the **same directory**.

Data Exporting

- **Save to text:** writetable, writematrix
- **Save to Excel:** writetable, xlswrite
- **Save to .mat:** save
- **Export figures:** saveas, exportgraphics

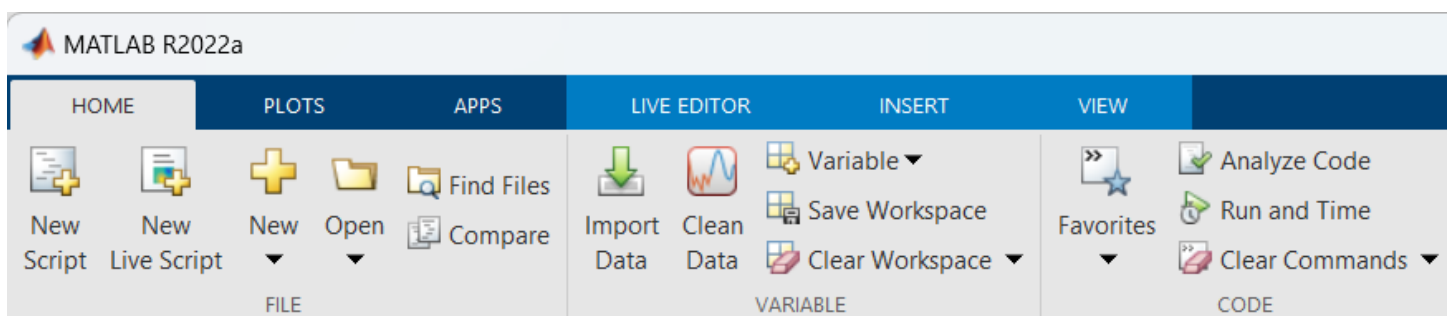
Image Data

```
im= imread('tsushima.png')    % reads the image file into MATLAB; saved in variable im  
  
imshow(im)                   % displays the image
```

- ⇒ The **'imread'** function allows you to read any image file into the workspace.
- ⇒ You need to include the extension (e.g., .jpeg, .png) with the filename.
- ⇒ The resulting data is a 3D array of size $1080 \times 1920 \times 3$, indicating a color image (also known as an **RGB image**) with three channels: **Red**, **Green**, and **Blue**. The third dimension corresponds to these color channels.
- ⇒ Each pixel value ranges from **0 to 255**, as the image is stored in **uint8** (8-bit unsigned integer) format. This allows **256 intensity levels** per channel, representing different shades of gray. This concept is often referred to as "**256 shades of white**."
- ⇒ The combination of values across the three channels at each pixel location determines the final color we perceive. With 256 possible values in each of the three channels, this allows for the representation of **over 16 million** distinct colors.

Excel File

You can read an Excel file into MATLAB using the **'readtable'** function or directly using the **'import data'** tab.



- ⇒ If you click on the import data tab, it will open a new window. You can choose your file if it is kept in that directory, or you can go to another location and import the file from there. A new window with the data file will open (image shown below). Click on 'import selection' to import the file into your workspace.
- ⇒ Another approach is to use the **'readtable'** function. You need to include the extension (.csv, .xlsx). Avoid using other functions like **'xlsread'** or **'readmatrix'**.

```
w= readtable('worldcities.csv')
```

- If you check the data, it is a 26562 x 11 **table**. This means there are 26,562 samples (cities) with 11 columns (also called attributes/features).
- The data contains both strings and numeric values. So, a table is a **heterogeneous data type**. We will learn more about this new data type in today's session.

The screenshot shows the MATLAB Import Wizard for 'worldcities.csv'. The 'Import' tab is active, showing 'Column delimiters: Comma', 'Range: A2:K265...', 'Output Type: Table', and 'unimportable cells with: NaN'. A button 'Import selected data into MATLAB Workspace' is visible. Below the wizard, the 'worldcities.csv' table is displayed with columns: city, city_ascii, lat, lng, country, iso2, iso3, admin_name, capital, population, and id. The table contains 30 rows of data, including cities like Tokyo, Jakarta, Delhi, Mumbai, Manila, Shanghai, São Paulo, Seoul, Mexico City, Guangzhou, Beijing, Cairo, New York, Kolkata, Moscow, Bangkok, Buenos Aires, Shenzhen, Dhaka, Lagos, Istanbul, Osaka, Karachi, Bangalore, Tehran, Kinshasa, Ho Chi Minh, Los Angeles, and Rio de Janeiro.

| | city | city_ascii | lat | lng | country | iso2 | iso3 | admin_name | capital | population | id |
|----|----------------|----------------|----------|-----------|----------------|------|------|----------------|---------|------------|------------|
| 1 | city | city_ascii | lat | lng | country | iso2 | iso3 | admin_name | capital | population | id |
| 2 | Tokyo | Tokyo | 35.6897 | 139.6922 | Japan | JP | JPN | Tokyo | primary | 37977000 | 1392685764 |
| 3 | Jakarta | Jakarta | -6.2146 | 106.8451 | Indonesia | ID | IDN | Jakarta | primary | 34540000 | 1360771077 |
| 4 | Delhi | Delhi | 28.6600 | 77.2300 | India | IN | IND | Delhi | admin | 29617000 | 1356872604 |
| 5 | Mumbai | Mumbai | 18.9667 | 72.8333 | India | IN | IND | Mahārāshtra | admin | 23355000 | 1356226629 |
| 6 | Manila | Manila | 14.5958 | 120.9772 | Philippines | PH | PHL | Manila | primary | 23088000 | 1608618140 |
| 7 | Shanghai | Shanghai | 31.1667 | 121.4667 | China | CN | CHN | Shanghai | admin | 22120000 | 1156073548 |
| 8 | São Paulo | Sao Paulo | -23.5504 | -46.6339 | Brazil | BR | BRA | São Paulo | admin | 22046000 | 1076532519 |
| 9 | Seoul | Seoul | 37.5833 | 127.0000 | Korea, South | KR | KOR | Seoul | primary | 21794000 | 1410836482 |
| 10 | Mexico City | Mexico City | 19.4333 | -99.1333 | Mexico | MX | MEX | Ciudad de ... | primary | 20996000 | 1484247881 |
| 11 | Guangzhou | Guangzhou | 23.1288 | 113.2590 | China | CN | CHN | Guangdong | admin | 20902000 | 1156237133 |
| 12 | Beijing | Beijing | 39.9050 | 116.3914 | China | CN | CHN | Beijing | primary | 19433000 | 1156228865 |
| 13 | Cairo | Cairo | 30.0561 | 31.2394 | Egypt | EG | EGY | Al Qāhirah | primary | 19372000 | 1818253931 |
| 14 | New York | New York | 40.6943 | -73.9249 | United States | US | USA | New York | admin | 18713220 | 1840034016 |
| 15 | Kolkata | Kolkata | 22.5411 | 88.3378 | India | IN | IND | West Bengal | admin | 17560000 | 1356060520 |
| 16 | Moscow | Moscow | 55.7558 | 37.6178 | Russia | RU | RUS | Moskva | primary | 17125000 | 1643318494 |
| 17 | Bangkok | Bangkok | 13.7500 | 100.5167 | Thailand | TH | THA | Krung Thep... | primary | 17066000 | 1764068610 |
| 18 | Buenos Aires | Buenos Aires | -34.5997 | -58.3819 | Argentina | AR | ARG | Buenos Aire... | primary | 16157000 | 1032717330 |
| 19 | Shenzhen | Shenzhen | 22.5350 | 114.0540 | China | CN | CHN | Guangdong | minor | 15929000 | 1156158707 |
| 20 | Dhaka | Dhaka | 23.7161 | 90.3961 | Bangladesh | BD | BGD | Dhaka | primary | 15443000 | 1050529279 |
| 21 | Lagos | Lagos | 6.4500 | 3.4000 | Nigeria | NG | NGA | Lagos | minor | 15279000 | 1566593751 |
| 22 | Istanbul | Istanbul | 41.0100 | 28.9603 | Turkey | TR | TUR | Istanbul | admin | 15154000 | 1792756324 |
| 23 | Osaka | Osaka | 34.6936 | 135.5019 | Japan | JP | JPN | Osaka | admin | 14977000 | 1392419823 |
| 24 | Karachi | Karachi | 24.8600 | 67.0100 | Pakistan | PK | PAK | Sindh | admin | 14835000 | 1586129469 |
| 25 | Bangalore | Bangalore | 12.9699 | 77.5980 | India | IN | IND | Karnāṭaka | admin | 13707000 | 1356410365 |
| 26 | Tehran | Tehran | 35.7000 | 51.4167 | Iran | IR | IRN | Tehrān | primary | 13633000 | 1364305026 |
| 27 | Kinshasa | Kinshasa | -4.3317 | 15.3139 | Congo (Kin...) | CD | COD | Kinshasa | primary | 13528000 | 1180000363 |
| 28 | Ho Chi Min... | Ho Chi Min... | 10.8167 | 106.6333 | Vietnam | VN | VNM | Hồ Chí Minh | admin | 13312000 | 1704774326 |
| 29 | Los Angeles | Los Angeles | 34.1139 | -118.4068 | United States | US | USA | California | admin | 12750807 | 1840020491 |
| 30 | Rio de Jane... | Rio de Jane... | -22.9083 | -43.1964 | Brazil | BR | BRA | Rio de Jane... | admin | 12272000 | 1076887657 |

- You can see from the above image that some of the columns, like 'id' or 'lat', are of numeric data type. While other columns like 'country' are of the '**Categorical**' data type.

Text files

You can read a .txt file using several functions depending on the file's structure and your requirements. Not every method is suitable for all files.

- If your file contains numeric data arranged in rows and columns, use '**readmatrix**'.
- Use '**fileread**' for text files where you need the full contents as a single string. It will read the entire file as a string.
- You can also use the '**import data**' tab to directly import into MATLAB. This will import the data using a table or array format.

There are two .txt files provided - 'Uni data.txt' and 'high impact conference.txt'. Both are of different types. The first one is similar to a .csv file with 'space' as a delimiter. The second one is a simple string.

Using all 3 methods mentioned above is not suitable – check the image below. It's for the first data.

- The first output is using the '**Import Data**' tab.
- The second output is from using the 'fileread' function – this reads the entire data as a single string, which does not serve the purpose of this data.
- The third output is from using the 'readmatrix' function – this reads the data only as numeric values. The strings are termed as NaN.

Evidently, the first one is most suitable for this data. Find out which one will be more suitable for the other file.

unidata = 7×6 table

| | VarName1 | BUET | RUET | KUET | CUET | I |
|---|----------------|------|------|------|------|---|
| 1 | "CSE" | 120 | 120 | 100 | 80 | |
| 2 | "EEE" | 180 | 120 | 150 | 80 | |
| 3 | "ME" | 150 | 80 | 150 | 80 | |
| 4 | "CIVIL" | 195 | 80 | 150 | 80 | |
| 5 | "CE" | 40 | NaN | 30 | 80 | |
| 6 | "Architecture" | 100 | 50 | 80 | 80 | |
| 7 | "Management" | 50 | 50 | NaN | 80 | |

t =

```

'    BUET    RUET    KUET    CUET    IUT
CSE      120    120    100    80    40
EEE      180    120    150    80    80
ME 150    80    150    80    55
CIVIL    195    80    150    80    45
CE 40     nan    30    80    nan
Architecture    100    50    80    80    nan
Management 50    50    nan    80    30
'
```

ta = 7×6

```

NaN    120    120    100    80    40
NaN    180    120    150    80    80
NaN    150    80    150    80    55
NaN    195    80    150    80    45
NaN    40     NaN    30    80    NaN
NaN    100    50    80    80    NaN
NaN    50     50    NaN    80    30
```

Mat files

.mat files are MATLAB's proprietary binary format used to store variables, arrays, and other data. They allow you to save and load your workspace variables efficiently between sessions or across programs. There are some sample .mat files already provided by MATLAB.

To read .mat files in MATLAB, you use the `load` function. This loads all variables stored in a .mat file into the workspace. You don't need to assign a separate variable.

```
load fatalities.mat
```

This will load a 49x8-sized table into the workspace under the variable name 'fatalities'.

Data Analysis/Processing

Data exploration, data cleaning, and data manipulation – all these can be considered part of data processing tasks. Let's look at some simple data processing tasks.

Image Data

Image processing is a field in Electrical and Computer Engineering that focuses on performing operations on images to analyze, enhance, or extract useful information from them. 'Digital Image Processing' is an elective course offered by the EEE department to 4th-year students. So, we will not explore much of this task in this course. If you are interested, you can look into the following GitHub repositories or YouTube to get some introduction (these will not be included in the assessment of this course).

- <https://github.com/newaz-aa/Digital-Image-Processing>
- https://github.com/newaz-aa/Biomedical_Image_Processing

Image processing includes tasks like removing noise, image enhancement, segmentation, feature extraction, object recognition, etc.

Some common operations on images include –

| Operation | Example |
|----------------------|--------------------------------|
| Grayscale conversion | RGB → black & white |
| Filtering | Blur, sharpen, denoise |
| Edge detection | Detect object boundaries |
| Morphological ops | Erosion, dilation, thinning |
| Image segmentation | Separate foreground/background |
| Object recognition | Identify faces, digits, etc. |

| Operation | Example |
|-------------|-----------------|
| Compression | JPEG, PNG, etc. |

Let's try one of them. We have imported an RGB image into MATLAB. Let's convert it into a grayscale and a binary image.

The grayscale image has the same width and height as the RGB image but with only 1 channel (values ranging from 0 to 255). The binary image also has the same width and height but with only logical values 0 and 1, representing black and white colors, respectively.

```
im_gray= rgb2gray(im)
imshow(im_gray)

im_bin= imbinarize(im_gray)
imshow(im_bin)

montage({im, im_gray, im_bin})
```



Applications of Image Processing:

- **Medical imaging:** MRI, CT scan analysis
- **Computer vision:** self-driving cars, facial recognition

- **Remote sensing:** satellite image analysis
- **Document analysis:** OCR: Optical Character Recognition
- **Security and surveillance:** Number plate recognition
- **Industrial automation:** defect detection, barcode reading

Tools for Image Processing:

- **MATLAB:** It provides an Image Processing Toolbox to perform different tasks. You can also find different built-in apps for 'image processing and computer vision' in the 'APPS' tab. These apps make different image-processing tasks much easier.
- **Python:** libraries like OpenCV, PIL (pillow), scikit-image, Pytesseract
- **C++/OpenCV**
- **TensorFlow/PyTorch:** for deep learning-based tasks

Text Data

Text data processing refers to the methods and techniques used to clean, transform, and analyze text data so it can be used in applications like natural language processing (NLP), sentiment analysis, and information retrieval. It primarily falls under the domain of Computer Science, so we will not explore this segment in depth.

MATLAB provides simple tools to handle text. For example, the 'fileread' function allows you to import text data directly into MATLAB as a string (character array). Once imported, you can perform various processing tasks.

The following example demonstrates how to create a **word cloud** from a text passage. A word cloud visually highlights the **most frequent words**, helping identify key terms in the passage.

```
war= fileread('War_2025.txt')  
wordcloud(war)
```

Applications of Text Processing

- Spam detection
- Sentiment analysis
- Chatbots / virtual assistants
- Machine translation
- Document classification
- Search engines

Table Data Structure

It is a data container that allows you to store columns of data of different types (e.g., numbers, strings, dates) all in one variable—perfect for structured data. It is similar to a spreadsheet or database table.

Key Characteristics of a table:

- Each column can have a different data type.
- Columns have names.
- Rows can have names (**optional**).
- You can access columns like fields in a structure.

Creating a Table

Suppose the following table contains the data of some students at IUT. How can we represent this data properly? The data contains numbers as well as strings. So, we need a heterogeneous data type.

All the students here have 4 **attributes** (ID, Gender, CGPA, Dep). There are 5 students in total – so 5 rows.

| Name | ID | Gender | CGPA | Department |
|---------------|-----|--------|------|------------|
| Ben Affleck | 112 | Male | 3.7 | EEE |
| Henry Cavil | 170 | Male | 3.8 | ME |
| Zack Snyder | 214 | Male | 3.9 | CSE |
| Hermione | 120 | Female | 3.85 | EEE |
| Lucy Pavensie | 220 | Female | 3.65 | CE |

Let's try to create this table in MATLAB.

- I. Start by creating several **column vectors** that contain your data. Each vector will represent a column in the table. These columns can contain different data types — like numbers, text, or logical values. For the above data, we need 5 such column vectors.

```
name= ["Ben Affleck"; "Henry Cavil"; "Zack Snyder"; "Hermione"; "Lucy Pavensie"]
Id= [112; 170; 214; 120; 220]
Gender = ["male"; "male"; "male"; "female"; "female"]
CGPA = [3.7, 3.8, 3.9, 3.85, 3.75]'
dept= ["EEE", "ME", "CSE", "EEE", "CE"]
```

There is one simple mistake in the above code. Can you figure out what that is?

- II. Then, combine these vectors into a single table using the `table()` function.

```
Data_v1= table(name, Id, Gender, CGPA, dept)
```

This will create a table of size (5, 5). As you can see, the ‘name’ variable is considered an attribute in the following table.

`data = 5x5 table`

| | name | Id | Gender | CGPA | dept |
|---|-----------------|-----|----------|--------|-------|
| 1 | "Ben Affleck" | 112 | "male" | 3.7000 | "EEE" |
| 2 | "Henry Cavil" | 170 | "male" | 3.8000 | "ME" |
| 3 | "Zack Snyder" | 214 | "male" | 3.9000 | "CSE" |
| 4 | "Hermione" | 120 | "female" | 3.8500 | "EEE" |
| 5 | "Lucy Pavensie" | 220 | "female" | 3.7500 | "CE" |

III. We can consider labeling each row of the table. That way, we can access the rows by that label, similar to column names.

Check the difference between the two tables – one is 5x5, the other is 5x4. Instead of using ‘name’, we could also have used ‘ID’ as the label for rownames.

```
data_v2= table(Id, Gender, CGPA, dept, RowNames=name)
```

`data_v2 = 5x4 table`

| | | Id | Gender | CGPA | dept |
|---|----------------------|-----|----------|--------|-------|
| 1 | Ben Affleck | 112 | "male" | 3.7000 | "EEE" |
| 2 | Henry Cavil | 170 | "male" | 3.8000 | "ME" |
| 3 | Zack Snyder | 214 | "male" | 3.9000 | "CSE" |
| 4 | Hermione | 120 | "female" | 3.8500 | "EEE" |
| 5 | Lucy Pavensie | 220 | "female" | 3.7500 | "CE" |

Accessing Data

You can access data from a table in several ways:

- By column name (like accessing fields in a structure).
- By using curly braces {} for numeric indexing to extract **data directly**.
- By using parentheses () to extract a **sub-table**.

```
data_v1.Gender

data_v1{2,:}
data_v1{2:4, "CGPA"}
data_v1{:[1, 3]}
data_v1{:[2, 4]}

data_v1(1:3, 5)
data_v2(:, 3:4)

data_v2("Ben Affleck", :)
data_v2{"Hermione", :}
```

Modifying Table

You can update existing data in the table by assigning new values to specific cells or columns. You can also:

- Add a new column by assigning a new vector to a new column name.
- Add a new row by using the same syntax and providing values for all columns.

- i. Let's say you want to change the ID of Ben Affleck to 200.

```
data_v1.Id(1)=200
```

- ii. Change the department of Lucy to ME.

```
data_v1.dept(5)='ME'
```

- iii. Change the CGPA of Cavil.

```
data_v2("Henry Cavil", "CGPA")= {3.9}
```

In this approach, you need to use { }.

- iv. Add a new column, age.

```
data_v1.age = [50, 45, 52, 24, 16]'
```

- v. Add a new row

```
data_v1(6,:) = {"Triss", 300, "female", 3.99, "EEE", 100} % hard coded
```

```
data_v1(size(data_v1,1),:)={"Geralt", 400, "male", 3.0, "CSE", 150}
```

Exporting Tables

You can save your modified table as a CSV or TXT file using the ‘writetable’ function. It will be saved in the current directory. Don’t forget to add the extension with the filename.

```
writetable(data_v1, 'data_modified.csv')
```

Data Processing/Analyzing Tables

You can perform various data analysis/processing tasks on the table. Here is a summary of some common functions for that.

| Function | Description |
|-----------------------|--|
| summary(T) | Summary of variables (useful for large tables) |
| head(T, n) | First n rows |
| tail(T, n) | Last n rows |
| sortrows(T) | Sort table rows |
| innerjoin / outerjoin | Join tables |
| removevars | Remove specific variables (columns) |
| renamevars | Rename columns |
| rowfun | Apply a function to each row |

- i. Say, you want to sort your table based on student ID or CGPA. You can use the ‘sortrows’ function with an optional argument indicating the column based on which you want to sort the rows.

```
sortrows(data_v1, "CGPA")
```

- ii. You can remove any column using the ‘removevars’ function.

```
removevars(data_v1, "dept")
```

- iii. You can obtain a summary of the data (max/min/avg value, variable type) using the summary function.

```
summary(data_v1)
```

Categorical Data Structure

MATLAB's categorical data type is used to store discrete, repeated text values more efficiently and meaningfully than plain strings or character arrays. It allows the **grouping** of similar strings.

For instance, you have labels "Male" and "Female" in the above table you created. The 'Gender' column is of type string (you can see from the output of the summary function). Let's change it to a categorical data type and see the difference.

```
data_v3= data_v1
data_v3.Gender = categorical (data_v1.Gender)
summary(data_v3)
```

Can you spot the difference between the summary of data_v1 and data_v3? The group of students has now been classified into two categories: male and female. They are no longer in string format.

This allows for easier handling in many cases. For instance, you can extract or **summarize** similar types of data much more efficiently when the data type for that variable is categorical.

| | name | Id | Gender | CGPA | dept | |
|---|-----------------|-----|----------|--------|-------|--|
| 1 | "Ben Affleck" | 200 | "male" | 3.7000 | "EEE" | |
| 2 | "Henry Cavil" | 170 | "male" | 3.8000 | "ME" | |
| 3 | "Zack Snyder" | 214 | "male" | 3.9000 | "CSE" | |
| 4 | "Hermione" | 120 | "female" | 3.8500 | "EEE" | |
| 5 | "Lucy Pavensie" | 220 | "female" | 3.7500 | "ME" | |
| 6 | "Geralt" | 400 | "male" | 3 | "CSE" | |

data_v3 = 6x6 table

| | name | Id | Gender | CGPA | dept | |
|---|-----------------|-----|--------|--------|-------|--|
| 1 | "Ben Affleck" | 200 | male | 3.7000 | "EEE" | |
| 2 | "Henry Cavil" | 170 | male | 3.8000 | "ME" | |
| 3 | "Zack Snyder" | 214 | male | 3.9000 | "CSE" | |
| 4 | "Hermione" | 120 | female | 3.8500 | "EEE" | |
| 5 | "Lucy Pavensie" | 220 | female | 3.7500 | "ME" | |
| 6 | "Geralt" | 400 | male | 3 | "CSE" | |

The benefit of this will become more apparent when you work with a huge amount of data. If you have a very large data set containing all the students of IUT from the beginning, categorized data can make it very easy to find a particular group, e.g., students of batch 2010 or the students of the BTM department, etc.

⇒ Try to plot the gender column of data_v1 and data_v3 using the **plots** tab. See the difference.

⇒ If you have millions of repeated strings, converting them to categorical can drastically **reduce memory usage**.

This is because categorical data stores only one instance of each category name, reducing the memory footprint.