

# CMPINF 2100

## Introduction to Data Centric Computing

Midterm Exam Data Overview

Problem overview, goals, and context

# So far this semester you have learned...

- Essential base Python programming
  - Data types, methods, attributes, for-loops, comprehensions, slicing, etc...
- Manipulation of NumPy arrays and Pandas DataFrames
  - Slicing – selecting columns and filtering rows
  - Concatenation
  - Reshaping
  - MERGE/JOIN DataFrames by column values
  - Summarize: mean, standard deviation, size, count, number of unique values
  - SPLIT-APPLY-COMBINE to summarize by GROUPS
- Data visualization to visually explore columns (variables) in DataFrames
  - Visualizations DEPEND on data type
  - Visualizations explore MARGINAL behavior (one variable at a time)
  - Visualizations explore CONDITIONAL behavior (group a variable by another)
- Cluster analysis and PCA to help find patterns in the data.

# You must use ALL aspects learned so far to explore a realistic data application

- You will work through an application very similar to many applications I worked on as a Data Scientist in the manufacturing industry.
- Multiple data sets are provided to you as CSV files.
  - midterm\_machine\_01.csv
  - midterm\_machine\_02.csv
  - midterm\_machine\_03.csv
  - midterm\_supplier.csv
  - midterm\_test.csv
- You will explore the variables within each file, JOIN them appropriately, and then explore the combined data to identify important patterns.

# The data provided to you is based on the following manufacturing scenario:

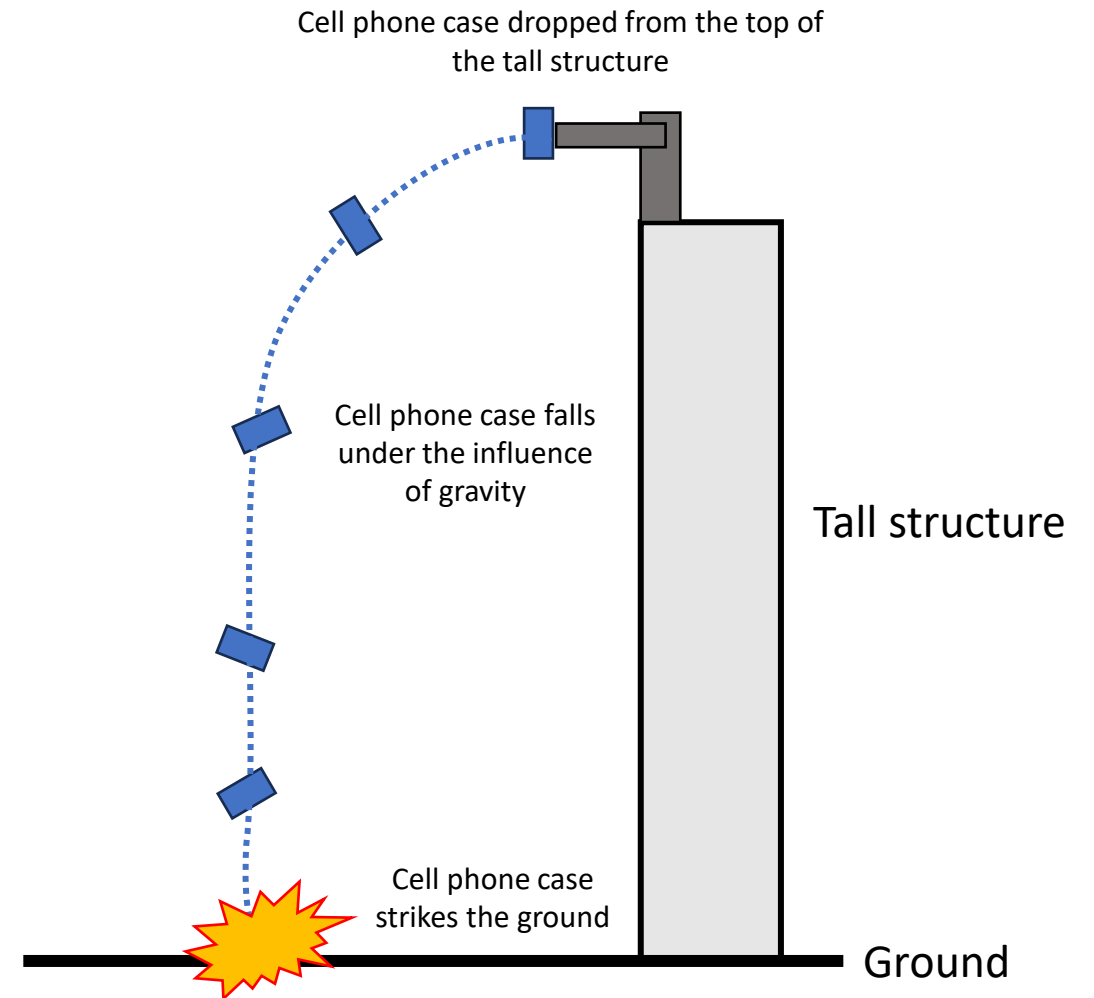
- A company manufactures cell phone cases. The cases are made from a high-density plastic.
- The company buys the plastic from **4 SUPPLIERS**.
  - The plastic is purchased in **BATCHES** from a **SUPPLIER**.
  - A single **BATCH** is represented by the plastic **DENSITY**.
- The cell phone cases are made with injection molding machines.
  - The company uses **3 MACHINES** to manufacture the cell phone cases.
  - Each **MACHINE** consists of **4 OPERATING VARIABLES** that define how the machine produces the cell phone case. The machines can be operated differently.
  - A single **BATCH** of plastic can be used across multiple **MACHINES**.

# The data provided to you is based on the following manufacturing scenario:

- The company wants to produce HIGH QUALITY cell phone cases. The cases are DROP TESTED to ensure they do NOT break or shatter under reasonable use.
  - If the cell phone case breaks during the DROP TEST, the test result is a FAIL.
- The company uses Data Science and Machine Learning techniques to examine if:
  - the DROP TEST FAILURE RATE varies across the plastic SUPPLIERS.
  - the DROP TEST FAILURE RATE varies across the MACHINES.
  - the OPERATING VARIABLES impact the DROP TEST FAILURE RATE.

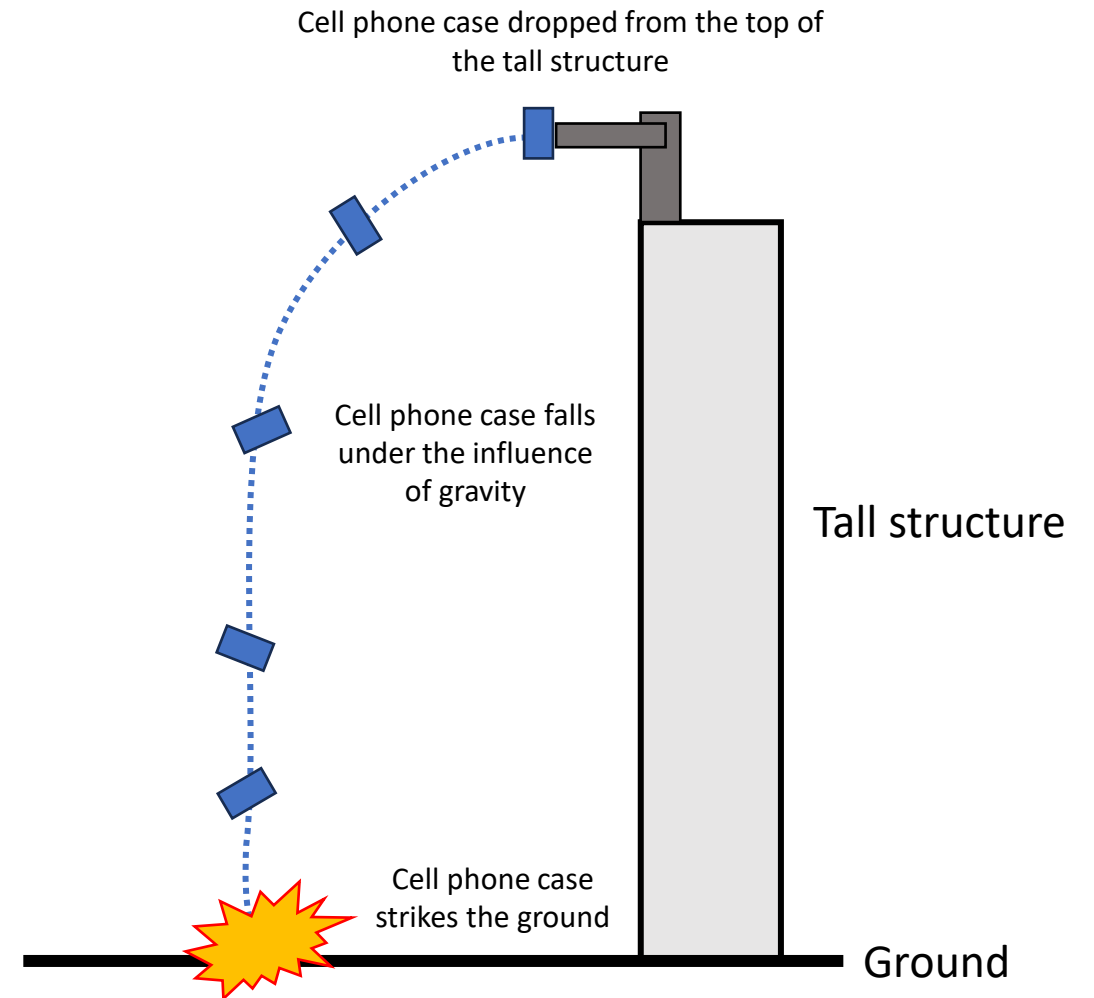
# DROP TEST overview

- The DROP TEST works by dropping a cell phone case with a representative phone inside it from the top of a tall structure.
- It falls under the influence of gravity until it strikes the ground.
- The case is inspected to see if it survived the fall and protected the phone.



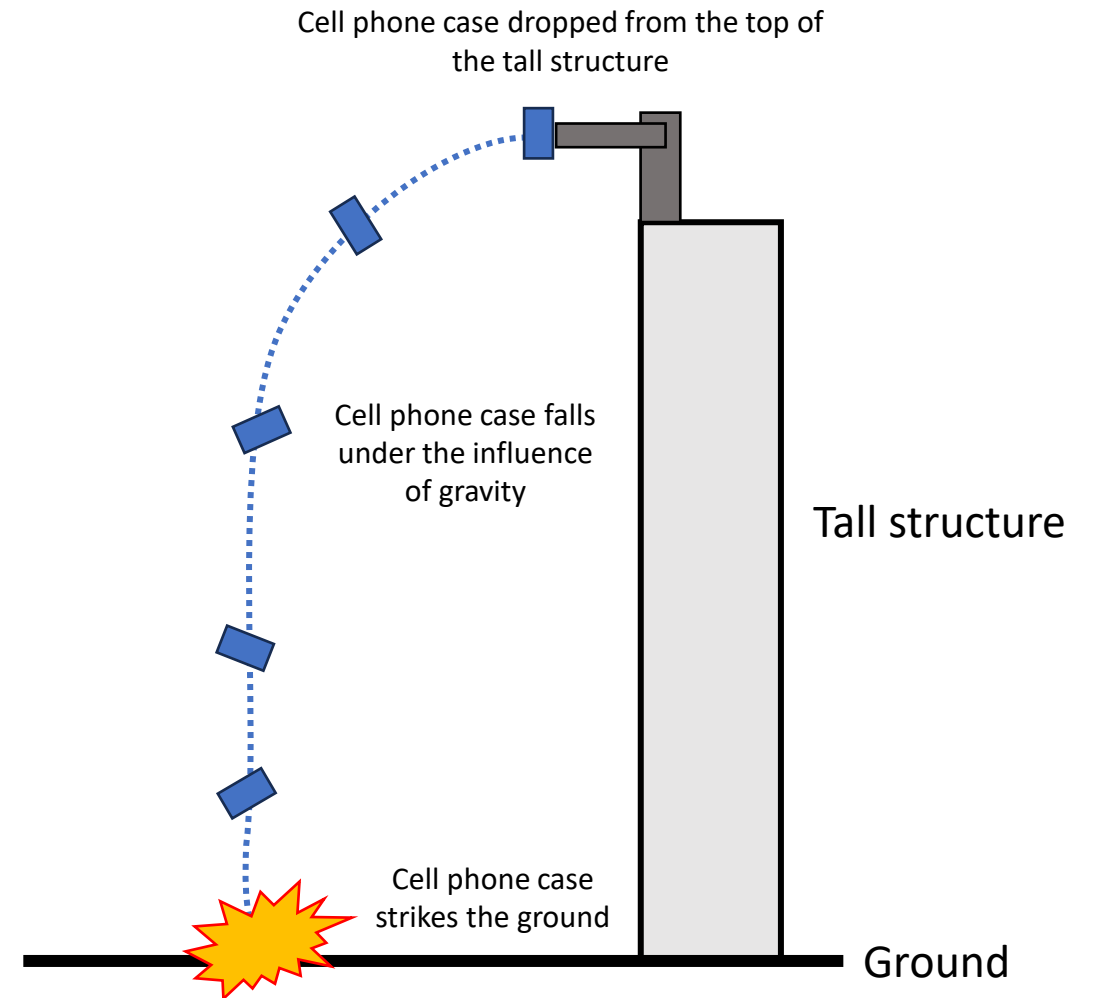
# DROP TEST overview

- The DROP TEST may destroy the cell phone case!
- Testing is also time consuming!
  - The company produces THOUSANDS of cases per day.
  - It would take a VERY LONG time to test every case.
- Thus, NOT all cases are tested!
- A SAMPLE of cases are collected and DROP TESTED. tested.



# DROP TEST overview

- The SAMPLING PLAN requires 10 out of every 100 cases manufactured per MACHINE to be DROP TESTED.
- The OPERATING VARIABLES used to produce the case on the MACHINE are recorded in a data base.
- The SUPPLIER that provided the plastic for each case is also recorded in a data base.





# The data provided to you is based on the following scenario:

- Ultimately, the TEST data are used to TRAIN a classifier to predict if a case will **FAIL** the drop test.
- The variables associated with the production of each case are known in a data base:
  - MACHINE, OPERATING VARIABLES, and SUPPLIER.
- Those variables are used as INPUTS to the classifier.
- However, you are NOT training the classifier in the midterm!
- You are EXPLORING the data. You will learn how to train classifiers AFTER the midterm.

# The manufacturing data are stored in separate CSV files for each machine

The MACHINE data are stored in 3 CSV files:

- midterm\_machine\_01.csv, midterm\_machine\_02.csv, midterm\_machine\_03.csv
- The file name tells you the machine ID the data come from:
  - For example, midterm\_machine\_01.csv is associated with Machine 1.

Each CSV file consists of 7 variables:

- ID: The unique unit ID for the cell phone case
- Batch: The batch index that denotes the plastic the cell phone case is created from
  - **NOTE:** the Batch is an INTEGER data type but is a CATEGORICAL variable.
- s\_id: The sequential production index for a single cell phone case within a Batch on a machine.
- 4 OPERATIONAL VARIABLES that describe the behavior of injection process: x1, x2, x3, and x4.

Machine 1

ID	Batch	s_id	x1	x2	x3	x4

Machine 2

ID	Batch	s_id	x1	x2	x3	x4

Machine 3

ID	Batch	s_id	x1	x2	x3	x4
					10	

# The batches of plastic material come from the 4 different suppliers

- **SUPPLIER** data are stored in the midterm\_supplier.csv file.
- That data table consists of 3 variables:
  - Batch: The batch index associated with the batch of plastic from the SUPPLIER
    - **NOTE**: the Batch is an INTEGER data type but is a CATEGORICAL variable.
  - Supplier: The supplier ID
  - Density: The supplier reported density associated with the batch of plastic

Batch	Supplier	Density

# The TEST results are stored in the midterm\_test.csv file

- That CSV file consists of 3 variables:
  - ID: The unique unit ID the tested cell phone case
  - test\_group\_id: Test grouping identification label
  - Result: The DROP TEST result which is encoded as:
    - A value of 1 corresponds to FAIL
    - A value of 0 corresponds to PASS

ID	test_group_id	Result

The 5 data tables need to be JOINED to ultimately link the INPUTS with the DROP TEST result.

ID	Batch	s_id	x1	x2	x3	x4
ID	Batch	s_id	x1	x2	x3	x4
ID	Batch	s_id	x1	x2	x3	x4

ID	test_group_id	Result

Batch	Supplier	Density

To do so we need to identify the common  
KEYS across the tables

ID	Batch	s_id	x1	x2	x3	x4
ID	Batch	s_id	x1	x2	x3	x4
ID	Batch	s_id	x1	x2	x3	x4

ID	test_group_id	Result

Batch	Supplier	Density

# And make sure we understand what one row represents in each data set!!!

ID	Batch	s_id	x1	x2	x3	x4
ID	Batch	s_id	x1	x2	x3	x4
ID	Batch	s_id	x1	x2	x3	x4

One row per **manufactured** cell case  
**per machine.**

Contains every manufactured cell  
phone case per machine.

ID	test_group_id	Result

One row per **tested** cell case.

Only contains test results for the  
tested cell phone cases.

Batch	Supplier	Density

One row per supplier provided  
**batch.**

# However, combining the machine data requires extra attention

ID	Batch	s_id	x1	x2	x3	x4

ID	Batch	s_id	x1	x2	x3	x4

ID	Batch	s_id	x1	x2	x3	x4

The machine tables do **NOT** contain any identifying information associated with the MACHINE!

The CSV file contains the machine ID!

You must **ADD** a column, machine\_id, that identifies the machine as 1, 2, or 3 BEFORE the machine data sets are combined.