

# MISSING VALUES HANDLING: CASE STUDY ON TSA CLAIMS DATABASE

by: Noor Kharismawan Akbar



# BACKGROUND

**Missing data** occur when no data values are stored for the variable under observation in statistics. This problem is quite common in many real-life datasets and can have a significant effect on the conclusions that can be drawn from it.

In this article, we will use the **TSA Claims Database** as a case study.

Dataset Link:

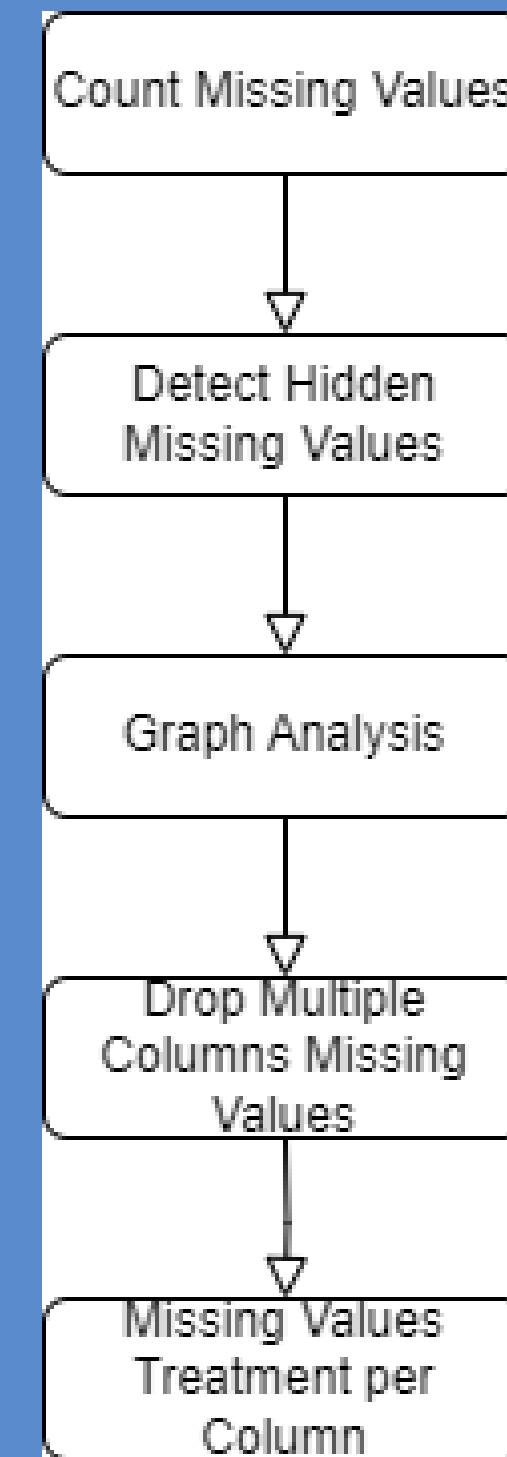
<https://www.kaggle.com/datasets/terminal-security-agency/tsa-claims-database>

	Claim Number	Date Received	Incident Date	Airport Code	Airport Name	Airline Name	Claim Type	Claim Site	Item	Claim Amount	Status	Close Amount	Disposition	
	4416	0401107L	1-Apr-03	2/20/2003 0:00	PHX	Phoenix Sky Harbor International	nan	nan	Checked Baggage	Locks	\$20.00	Approved	\$20.00	Approve in Full
	4481	0401004L	1-Apr-03	11/29/2002 0:00	RSW	Southwest Florida International	nan	Property Damage	Checkpoint	Cameras - Digital	\$120.00	Settled	\$40.00	Settled
	4479	0401002L	1-Apr-03	2/16/2003 0:00	LGA	LaGuardia	nan	Property Damage	Checked Baggage	Clothing - Shoes; belts; accessories; etc.	\$550.00	Settled	\$275.00	Settled
	4478	0401128L	1-Apr-03											
	4477	0401127L	1-Apr-03											

DF.NAN

# MISSING VALUES HANDLING FLOW

Following are the steps for handling missing values in this case study.



**FLOWCHART**

# MISSING VALUES CHECK

- The dataset consists of **204,267 rows & 13 columns**. It can also be seen that there are several missing values. All existing data types are in the form of object.
- It can be seen that the missing values have the **largest** value at **~30%** for Close Amount & Disposition values. If a variable has **less than 50%** missing values, **imputation** might be a viable option.

RangeIndex: 204267 entries, 0 to 204266

Data columns (total 13 columns):

#	Column	Non-Null Count	Dtype
0	Claim Number	204267 non-null	object
1	Date Received	204004 non-null	object
2	Incident Date	202084 non-null	object
3	Airport Code	195743 non-null	object
4	Airport Name	195743 non-null	object
5	Airline Name	169893 non-null	object
6	Claim Type	196354 non-null	object
7	Claim Site	203527 non-null	object
8	Item	200301 non-null	object
9	Claim Amount	200224 non-null	object
10	Status	204262 non-null	object
11	Close Amount	135315 non-null	object
12	Disposition	131359 non-null	object

dtypes: object(13)

	index	Total Null Values	Percentage
0	Disposition	72885	35.685259
1	Close Amount	68929	33.748360
2	Airline Name	34373	16.829381
3	Airport Code	8523	4.172950
4	Airport Name	8523	4.172950
5	Claim Type	7913	3.874288
6	Claim Amount	4043	1.979495
7	Item	3966	1.941795
8	Incident Date	2183	1.068820
9	Claim Site	740	0.362312
10	Date Received	263	0.128768
11	Status	5	0.002448
12	Claim Number	0	0.000000

DF.INFO

NULL VAL CHECK

	Claim Number	Date Received	Incident Date	Airport Code	Airport Name	Airline Name	Claim Type	Claim Site	Item	Claim Amount	Status	Close Amount	Disposition
182896	2013090606615	1-Aug-13	6/24/2013 0:00	-	-	-	Passenger Property Loss	-	-	-	-	nan	nan
182930	2013091006736	1-Aug-13	7/27/2013 0:00	-	-	Delta Air Lines	Passenger Property Loss	Checked Baggage	Other	-	-	nan	nan
157158	2011030180629	1-Feb-11	2/1/2011 0:00	-	-	-	Passenger Property Loss	Checked Baggage	Other	-	-	nan	nan
168054	2012030791604	1-Feb-12	9/13/2011 0:00	-	-	-	Passenger Property Loss	-	-	-	-	nan	nan
161706	2011071585243	1-Jul-11	6/28/2011 20:40	-	-	UAL	Passenger Property Loss	Checked Baggage	Cosmetics & Grooming	-	-	nan	nan

HIDDEN MISS. VAL

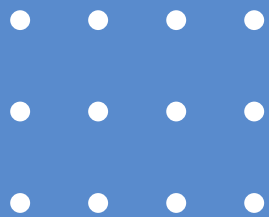


	Claim Number	Date Received	Incident Date	Airport Code	Airport Name	Airline Name	Claim Type	Claim Site	Item	Claim Amount	Status	Close Amount	Disposition
182896	2013090606615	1-Aug-13	6/24/2013 0:00	nan	nan	nan	Passenger Property Loss	nan	nan	nan	nan	nan	nan
182930	2013091006736	1-Aug-13	7/27/2013 0:00	nan	nan	Delta Air Lines	Passenger Property Loss	Checked Baggage	Other	nan	nan	nan	nan
157158	2011030180629	1-Feb-11	2/1/2011 0:00	nan	nan	nan	Passenger Property Loss	Checked Baggage	Other	nan	nan	nan	nan
168054	2012030791604	1-Feb-12	9/13/2011 0:00	nan	nan	nan	Passenger Property Loss	nan	nan	nan	nan	nan	nan
161706	2011071585243	1-Jul-11	6/28/2011 20:40	nan	nan	UAL	Passenger Property Loss	Checked Baggage	Cosmetics & Grooming	nan	nan	nan	nan

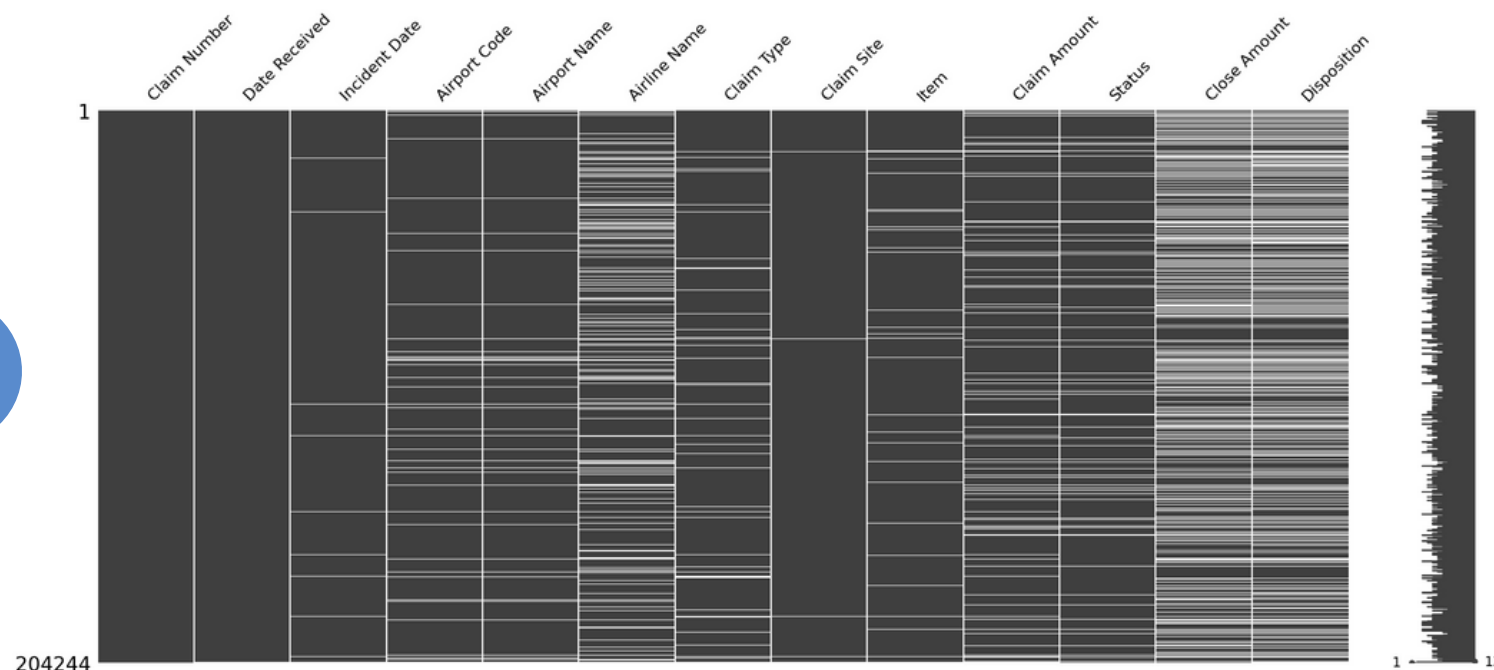
NP.NAN MISS. VAL

# HIDDEN MISSING VALUES

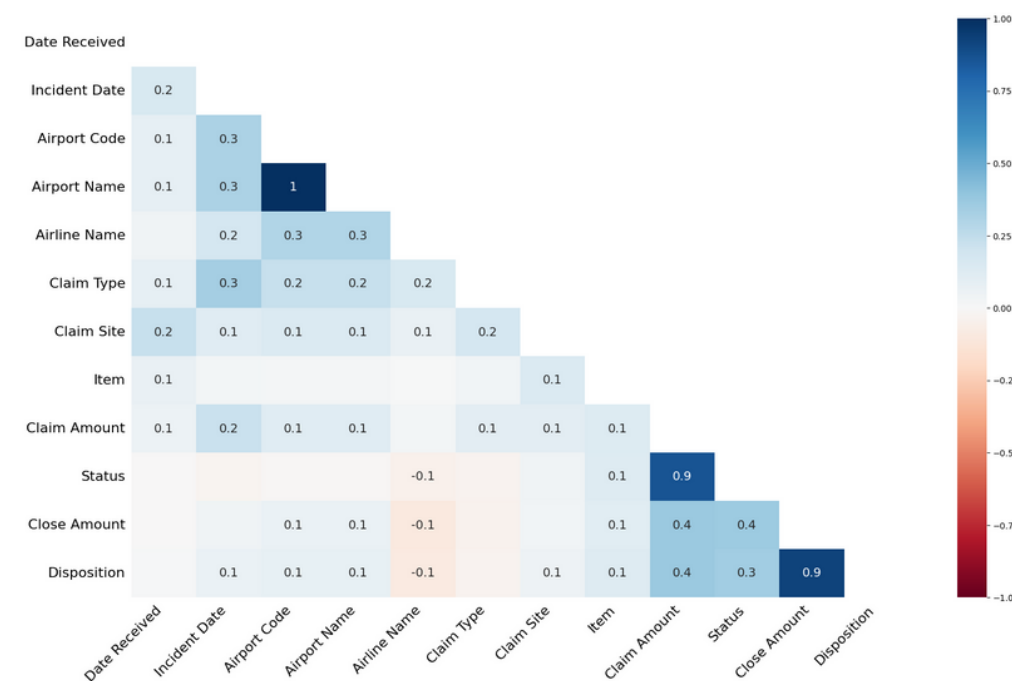
- There are some missing values, we will handle it. But, the initial step is identifying **hidden missing values (that is '-')**.
- That type of missing values must be **changed to numpy missing values** using np.nan.



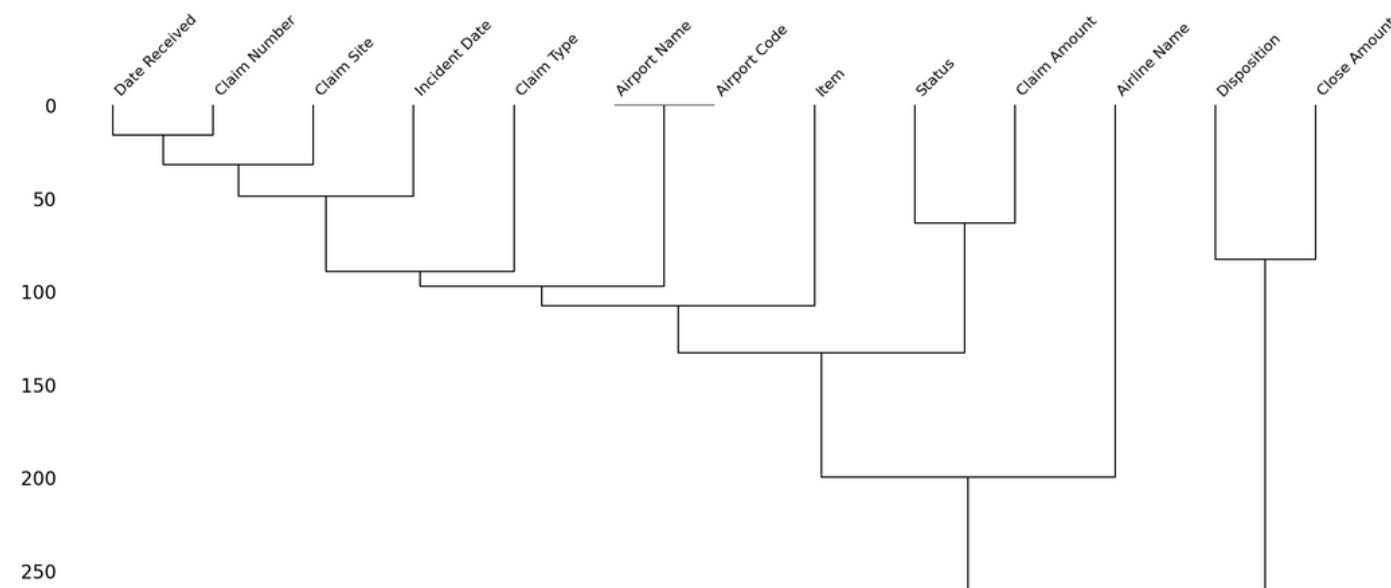
## MATRIX



## HEATMAP



## DENDOGRAM



# GRAPH ANALYSIS

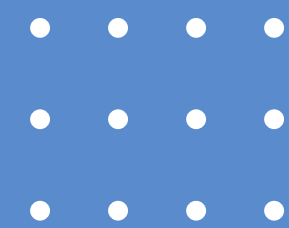
Missing values analysis:

1. Missing values are **scattered**
2. In the **same row**, there are also **several empty columns**
3. The **Disposition**, **Close Amount**, and **Airline Name** columns have many missing values
4. **[Status-Claim Amount]** & **[Close Amount-Disposition]** column has a high missing value correlation (**0.9**).
5. The **Airport Code & Airport Name** column has a missing value correlation with a value of **1**, whereas the two columns have a missing value with the **exact same number & and location**
6. Other columns have a **fairly low correlation** between columns

Because it only contains a **little information** and if inputting is done it will cause a **lot of bias**.

	Claim Number	Date Received	Incident Date	Airport Code	Airport Name	Airline Name	Claim Type	Claim Site	Item	Claim Amount		Status	Close Amount	Disposition
31941	2004051452119	1-Apr-04	3/9/2004 0:00	NaN 1	NaN 2	NaN 3	NaN 4	NaN 5	NaN 6	NaN 7		Insufficient; one of the following items required: sum certain; statement of fact; signature; location of incident; and date.	NaN 8	NaN 9
182896	2013090606615	1-Aug-13	6/24/2013 0:00	NaN 1	NaN 2	NaN 3	Passenger Property Loss	NaN 4	NaN 5	NaN 6		NaN 7	NaN 8	NaN 9
21287	2004072059830	1-Dec-03	11/4/2003 0:00	NaN 1	NaN 2	NaN 3	NaN 4	NaN 5	NaN 6	NaN 7		Canceled	NaN 8	NaN 9
168054	2012030791604	1-Feb-12	9/13/2011 0:00	NaN 1	NaN 2	NaN 3	Passenger Property Loss	NaN 4	NaN 5	NaN 6		NaN 7	NaN 8	NaN 9
149237	2010061572677	1-Jun-10	5/30/2010 0:00	NaN 1	NaN 2	NaN 3	NaN 4	NaN 5	NaN 6	NaN 7		NaN 8	NaN 9	NaN 10

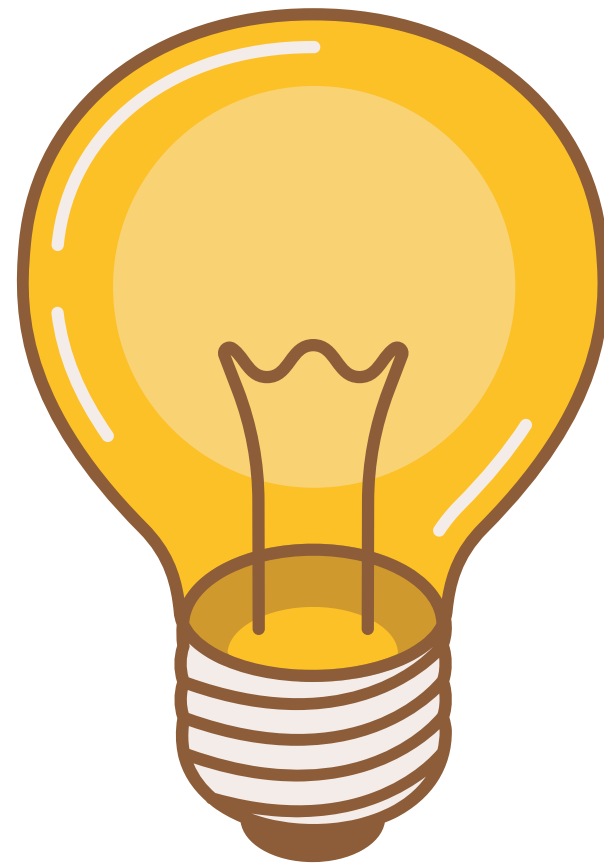
## MULTIPLE MISSING VALUES





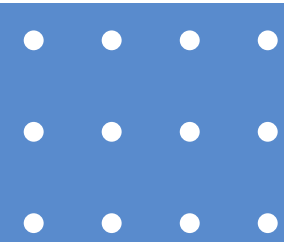






## WRAP UP

- There are many ways to deal with missing values.
- The first important thing is we need to master the domain knowledge.
- Then, we must look at the many types of missing values.
- After that, we must categorize the missing values and we can do many alternatives such as dropping, inputting, or keeping it blank.
- The Most important thing to consider is the treatment of the missing value must not make the data biased.



# THANK YOU !

**CONTACT ME!**



+6281227223150



akbar.noorkharismawan@gmail.com



<http://www.linkedin.com/in/n-k-akbar>



<https://github.com/baramizzo58>



<https://public.tableau.com/app/profile/akbar2070>

