

```
import pandas as pd
import numpy as np
from google.colab import files
```

```
uploaded = files.upload()
```



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Saving amazon-sales.csv to amazon-sales.csv

```
df = pd.read_csv('amazon-sales.csv')
```



/tmp/ipython-input-4-139105417.py:1: DtypeWarning: Columns (23) have mixed
df = pd.read_csv('amazon-sales.csv')

```
print(df.shape)
```



(128975, 24)


df.head(-10)



	index	Order ID	Date	Status	Fulfilment	Sales Channel	ship-service-level	Sty
0	0	405-8078784-5731545	04-30-22	Cancelled	Merchant	Amazon.in	Standard	SET3
1	1	171-9198151-1101146	04-30-22	Shipped - Delivered to Buyer	Merchant	Amazon.in	Standard	JNE37
2	2	404-0687676-7273146	04-30-22	Shipped	Amazon	Amazon.in	Expedited	JNE33
3	3	403-9615377-8133951	04-30-22	Cancelled	Merchant	Amazon.in	Standard	J03
4	4	407-1069790-7240320	04-30-22	Shipped	Amazon	Amazon.in	Expedited	JNE36
...
128960	128960	402-0468123-8401109	05-31-22	Shipped	Amazon	Amazon.in	Expedited	J03
128961	128961	402-0082204-6323568	05-31-22	Cancelled	Amazon	Amazon.in	Expedited	JNE37
128962	128962	408-9803724-6565965	05-31-22	Cancelled	Amazon	Amazon.in	Expedited	MEN50
128963	128963	404-5963451-7335564	05-31-22	Shipped	Amazon	Amazon.in	Expedited	J03
128964	128964	404-2225394-8024308	05-31-22	Shipped	Amazon	Amazon.in	Expedited	J01

128965 rows x 24 columns

```
df = df.drop(columns=['Unnamed: 22'])
df.head(5)
```



	index	Order ID	Date	Status	Fulfilment	Sales Channel	ship-service-level	Style	
0	0	405-8078784-5731545	04-30-22	Cancelled	Merchant	Amazon.in	Standard	SET389	SK
1	1	171-9198151-1101146	04-30-22	Shipped - Delivered to Buyer	Merchant	Amazon.in	Standard	JNE3781	JNKF
2	2	404-0687676-7273146	04-30-22	Shipped	Amazon	Amazon.in	Expedited	JNE3371	JN
3	3	403-9615377-8133951	04-30-22	Cancelled	Merchant	Amazon.in	Standard	J0341	
4	4	407-1069790-7240320	04-30-22	Shipped	Amazon	Amazon.in	Expedited	JNE3671	JNTL

5 rows x 23 columns

```
print(df.isnull().sum())
```

```

↗ index          0
  Order ID       0
   Date         0
  Status        0
 Fulfilment     0
 Sales Channel  0
ship-service-level 0
  Style         0
   SKU          0
  Category      0
   Size         0
  ASIN          0
 Courier Status 6872
   Qty          0
 currency      7795
 Amount       7795
ship-city       33
ship-state      33
ship-postal-code 33
ship-country    33
promotion-ids  49153
   B2B          0
fulfilled-by    89698
dtype: int64

```

```
df['fulfilled-by'].value_counts()
```

```

↗

```

	count
fulfilled-by	
Easy Ship	39277

```

dtype: int64

```

As here all the shipment is fulfilled by "Easy Ship", there is no harm in assuming that the missing shipment duty must have been assigned to the same company. Whether or not the shipment was delivered, it must have been assigned to "Easy Ship". We will fill those missing values with "Easy Ship"

```
df['fulfilled-by'] = df['fulfilled-by'].fillna('Easy Ship')
df['fulfilled-by'].value_counts()
```

↗

	count
--	-------

fulfilled-by

Easy Ship	128975
-----------	--------

dtype: int64

```
print(df.isnull().sum())
```

↗

index	0
Order ID	0
Date	0
Status	0
Fulfilment	0
Sales Channel	0
ship-service-level	0
Style	0
SKU	0
Category	0
Size	0
ASIN	0
Courier Status	6872
Qty	0
currency	7795
Amount	7795
ship-city	33
ship-state	33
ship-postal-code	33
ship-country	33
promotion-ids	49153
B2B	0
fulfilled-by	0
dtype: int64	

```
alignment = df['currency'].isnull() == df['Amount'].isnull()
all_aligned = alignment.all()
print(all_aligned)
```

↗ True

This above code ensures that entries found missing in the "currency" and "Amount" columns match exactly. Means the missing values are matched pair wise.

I suspected that because the number of missing values matched exactly.

Let's check if that's the case with the other four columns where missing values match!

```
new_alignment = (df['ship-city'].isnull() == df['ship-state'].isnull()) & \
                 (df['ship-state'].isnull() == df['ship-postal-code'].isnull())
                 (df['ship-postal-code'].isnull() == df['ship-country'].isnull()
new_all_aligned = new_alignment.all()
print(new_all_aligned)
```

➞ True

They match!

Below we are dropping Courier Status and Promotion Ids columns as they are not important for our analysis in this project

```
df = df.drop(columns=['Courier Status', 'promotion-ids'])
```

```
mask_col2_missing = df['ship-city'].isnull()
```

```
mask_col1_missing = df['currency'].isnull()
```

```
all_col2_missing_in_col1 = (mask_col1_missing[mask_col2_missing]).all()
print(all_col2_missing_in_col1)
```

➞ False

Here we got to know that the missing values form the two groups, where the missing values matched in rows, do not match.

We are going to drop all these messing data rows

We are doing this as it's impact on our analysis would be hardly more than 6%

We can work with that buffer.

```
df.dropna(subset=['currency'], inplace=True)
df.dropna(subset=['ship-city'], inplace=True)
print(df.isnull().sum())
```

```

↗ index                0
  Order ID              0
   Date                0
  Status              0
 Fulfilment            0
Sales Channel          0
ship-service-level    0
  Style                0
   SKU                0
  Category            0
   Size              0
  ASIN                0
   Qty                0
 currency            0
  Amount            0
ship-city            0
ship-state          0
ship-postal-code    0
ship-country        0
   B2B                0
fulfilled-by        0
dtype: int64
```

Hereby, we have successfully cleaned the data from Amazon Sales with a loss of 6% of data only.

This dataframe is now ready to be utilized for building a report.

```
print(df.shape)
```

```
↗ (121149, 21)
```

```
df.to_csv('cleaned_data.csv', index=False)
files.download('cleaned_data.csv')
```

```
↗
```