

```
# Cleaning and preprocessing of data.
import numpy as np
import pandas as pd
```

```
df = pd.read_csv('products (2).csv')
```

```
print(df)
```

```

      Index      Name \
0         1  Smart Fan Iron Cooker Go Wireless Portable
1         2                               Fan
2         3  Smart Speakerphone Charger Eco Plus Clean
3         4      Premium Grill Trimmer Portable
4         5      Keyboard Freezer
...      ...
9995      9996      Powerbank
9996      9997  Rechargeable Lamp Scanner Vacuum
9997      9998      Heater
9998      9999  Rechargeable Headphones Stove
9999     10000  Clean Charger Treadmill

      Description \
0      Catch enough role nearly.
1      All movement yeah tax me.
2      Quickly inside pull line lay start.
3      Lawyer one than fire.
4      Remain Congress blood plan voice.
...      ...
9995      Yeah none boy fund city decision.
9996  Appear country enter there pull thing computer.
9997  Owner plan approach soldier church little little.
9998      Base account partner probably.
9999      Return reflect truth tell scientist season.

      Brand      Category  MRP_Price  Currency  Stock \
0      Herman Ltd  Kids' Clothing    585.0      INR    194.0
1  Braun, King and Rollins  Grooming Tools    992.0      INR    724.0
2      Peck-Coleman  Fishing & Hunting    940.0      INR    769.0
3      Hines Ltd  Skincare    324.0      INR     93.0
4  Spence, Webster and Orr  Laptops & Computers    908.0      INR    614.0
...      ...
9995  Castaneda-Harrison  Books & Stationery    567.0      INR     27.0
9996      Knapp-Key  Cycling    300.0      INR    177.0
9997  Meyer, Case and Mills  Fitness Equipment    399.0      INR    996.0
9998      Jones-Bell  Kitchen Appliances     45.0      INR    274.0
9999      Davis LLC  Sports & Outdoors     31.0      INR     46.0

      EAN      Color  Availability  Internal ID  Qty \
0  3.968600e+12  Cornsilk  limited_stock      54  548.0
1  1.911270e+11  Bisque  discontinued      49  228.0
2  7.569140e+12  Blue  pre_order      42  520.0
3  2.705140e+12  Ivory  out_of_stock      93  251.0
4  9.830390e+12  FloralWhite  discontinued      91  602.0
...      ...
9995  6.519000e+12  Cyan  pre_order       7  206.0
9996  4.707960e+12  MintCream  limited_stock      55  557.0
9997  5.286510e+12  LightGray  pre_order      37  698.0
9998  3.702650e+12  SlateBlue  in_stock      52  183.0
9999  8.879840e+12  Ivory  backorder      70  904.0

      Selling_price
0      345.15
1      585.28
2      601.60
3      178.20
4      572.04
```

```
# check how much table and column
df.shape
```

```
(10000, 14)
```

```
#check type data form each column
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 14 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Index      10000 non-null  int64
1   Name       10000 non-null  object
2   Description 10000 non-null  object
3   Brand      10000 non-null  object
```

```
4  Category      10000 non-null  object
5  MRP_Price     9997 non-null   float64
6  Currency      10000 non-null  object
7  Stock         9978 non-null   float64
8  EAN           10000 non-null  float64
9  Color         10000 non-null  object
10 Availability  10000 non-null  object
11 Internal ID   10000 non-null  int64
12 Qty          9991 non-null   float64
13 Selling_price 9985 non-null   float64
dtypes: float64(5), int64(2), object(7)
memory usage: 1.1+ MB
```

```
#check null data
df.isna().any()
```

	0
Index	False
Name	False
Description	False
Brand	False
Category	False
MRP_Price	True
Currency	False
Stock	True
EAN	False
Color	False
Availability	False
Internal ID	False
Qty	True
Selling_price	True

dtype: bool

```
#add up the blank values
df.isna().sum()
```

	0
Index	0
Name	0
Description	0
Brand	0
Category	0
MRP_Price	3
Currency	0
Stock	22
EAN	0
Color	0
Availability	0
Internal ID	0
Qty	9
Selling_price	15

dtype: int64

```
#check duplicate data
df.duplicated().sum()
```

```
np.int64(0)
```

```
#check unique value
df.nunique()
```

	0
Index	10000
Name	7896
Description	10000
Brand	9241
Category	34
MRP_Price	999
Currency	1
Stock	999
EAN	9952
Color	140
Availability	6
Internal ID	99
Qty	951
Selling_price	7905

dtype: int64

```
#descirbe all
df.describe(include="all").T
```

	count	unique	top	freq	mean	std	min	25%
Index	10000.0	NaN	NaN	NaN	5000.5	2886.89568	1.0	2500.75
Name	10000	7896	Fan	44	NaN	NaN	NaN	NaN
Description	10000	10000	Between key father half.	1	NaN	NaN	NaN	NaN
Brand	10000	9241	Kennedy LLC	5	NaN	NaN	NaN	NaN
Category	10000	34	Clothing & Apparel	320	NaN	NaN	NaN	NaN
MRP_Price	9997.0	NaN	NaN	NaN	503.327498	289.746061	1.0	251.0
Currency	10000	1	INR	10000	NaN	NaN	NaN	NaN
Stock	9978.0	NaN	NaN	NaN	499.14071	288.02384	1.0	249.0
EAN	10000.0	NaN	NaN	NaN	5035709017000.380859	2887653409184.608887	261275896.0	2545170000000.0
Color	10000	140	Cornsilk	101	NaN	NaN	NaN	NaN
Availability	10000	6	discontinued	1706	NaN	NaN	NaN	NaN
Internal ID	10000.0	NaN	NaN	NaN	50.4755	28.555993	1.0	26.0
Qty	9991.0	NaN	NaN	NaN	523.69983	276.426889	50.0	281.5

```
df.isna().sum()
```

```

      0
  _____
  Index      0
  Name       0
  Description 0
  Brand       0
  Category    0
  MRP_Price   3
  Currency    0
  Stock      22
  EAN         0
  Color       0
  Availability 0
  Internal ID 0
  Qty         9
  Selling_price 15

```

```
dtype: int64
```

```
#fill na with 0
df['Stock'].fillna(0,inplace=True)
```

/tmp/ipython-input-2987266135.py:2: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through ch...
The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are...
For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col]

```
df['Stock'].fillna(0,inplace=True)
```

```
#fill na with 0

df['Qty'].fillna(0,inplace=True)
```

/tmp/ipython-input-290176259.py:1: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through ch...
The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are...
For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col]

```
df['Qty'].fillna(0,inplace=True)
```

```
# blank mrp_price analysis
```

```
mean_mrp_price = df['MRP_Price'].mean().round(2)
```

```
print(mean_mrp_price)
```

```
503.33
```

```
median_mrp_price = df['MRP_Price'].median()
print(median_mrp_price)
```

```
506.0
```

```
# brand wise calculating mean, median for mrp_price
```

```
brand_mrp_stats = df.groupby("Brand")["MRP_Price"].agg(
    mean_mrp="mean",
    median_mrp="median",
    count="count"
).reset_index()

print(brand_mrp_stats)
```

```

      0      Brand  mean_mrp  median_mrp  count
0      Abbott Ltd      95.0        95.0      1
1  Abbott, Holden and Conrad  800.0       800.0      1
2  Abbott, Lynch and Warner  521.0       521.0      1

```

```
3      Abbott, Pitts and Charles    527.0    527.0    1
4      Abbott, Zavala and Patrick   611.0    611.0    1
...
9236   Zuniga, Schwartz and Barber  664.0    664.0    1
9237   Zuniga, Zavala and Contreras  71.0     71.0     1
9238           Zuniga-Allen         496.0    496.0    1
9239           Zuniga-Mcpherson      969.0    969.0    1
9240           Zuniga-Yates          32.0     32.0     1
```

```
[9241 rows x 4 columns]
```

```
# Missing MRP values were filled using brand-wise median pricing, with overall median as fallback where brand-level data was missing
```

```
mrp_median = df.groupby("Brand")["MRP_Price"].median()
print(mrp_median)
```

```
Brand
Abbott Ltd          95.0
Abbott, Holden and Conrad  800.0
Abbott, Lynch and Warner   521.0
Abbott, Pitts and Charles  527.0
Abbott, Zavala and Patrick  611.0
...
Zuniga, Schwartz and Barber  664.0
Zuniga, Zavala and Contreras  71.0
Zuniga-Allen          496.0
Zuniga-Mcpherson       969.0
Zuniga-Yates           32.0
Name: MRP_Price, Length: 9241, dtype: float64
```

```
df["MRP_Price"] = df["MRP_Price"].fillna(
    df["Brand"].map(mrp_median)
)
```

```
overall_median = df["MRP_Price"].median()
df["MRP_Price"] = df["MRP_Price"].fillna(overall_median)
```

```
print(overall_median)
```

```
506.0
```

```
# selling price analysis
```

```
mean_selling_price = df['Selling_price'].mean()
```

```
print(mean_selling_price)
```

```
365.3695943915874
```

```
median_selling_price = df['Selling_price'].median()
```

```
print(median_selling_price)
```

```
353.4
```

```
# brand wise calculating mean, median for selling price
```

```
brand_selling_stats = df.groupby("Brand")["Selling_price"].agg(
    mean_mrp="mean",
    median_mrp="median",
    count="count"
).reset_index()

print(brand_selling_stats)
```

```
      Brand  mean_mrp  median_mrp  count
0  Abbott Ltd      60.80      60.80     1
1  Abbott, Holden and Conrad  488.00  488.00     1
2  Abbott, Lynch and Warner   401.17  401.17     1
3  Abbott, Pitts and Charles   447.95  447.95     1
4  Abbott, Zavala and Patrick   446.03  446.03     1
...
9236  Zuniga, Schwartz and Barber  544.48  544.48     1
9237  Zuniga, Zavala and Contreras    46.86    46.86     1
9238           Zuniga-Allen      381.92   381.92     1
9239           Zuniga-Mcpherson   629.85   629.85     1
9240           Zuniga-Yates     16.64    16.64     1
```

```
[9241 rows x 4 columns]
```

```
overall_selling_price_median = df["Selling_price"].median()
df["Selling_price"] = df["Selling_price"].fillna(overall_selling_price_median)
```

```
print(overall_selling_price_median)
```

353.4

```
df.isna().sum()
```

	0
Index	0
Name	0
Description	0
Brand	0
Category	0
MRP_Price	0
Currency	0
Stock	0
EAN	0
Color	0
Availability	0
Internal ID	0
Qty	0
Selling_price	0

dtype: int64

```
df
```

	Index	Name	Description	Brand	Category	MRP_Price	Currency	Stock	EAN	Color	Availability
0	1	Smart Fan Iron Cooker Go Wireless Portable	Catch enough role nearly.	Herman Ltd	Kids' Clothing	585.0	INR	194.0	3.968600e+12	Cornsilk	limited_st
1	2	Fan	All movement yeah tax me.	Braun, King and Rollins	Grooming Tools	992.0	INR	724.0	1.911270e+11	Bisque	discontin
2	3	Smart Speakerphone Charger Eco Plus Clean	Quickly inside pull line lay start.	Peck- Coleman	Fishing & Hunting	940.0	INR	769.0	7.569140e+12	Blue	pre_or
3	4	Premium Grill Trimmer Portable	Lawyer one than fire.	Hines Ltd	Skincare	324.0	INR	93.0	2.705140e+12	Ivory	out_of_st
4	5	Keyboard Freezer	Remain Congress blood plan voice.	Spence, Webster and Orr	Laptops & Computers	908.0	INR	614.0	9.830390e+12	FloralWhite	discontin
...
9995	9996	Powerbank	Yeah none boy fund city decision.	Castaneda- Harrison	Books & Stationery	567.0	INR	27.0	6.519000e+12	Cyan	pre_or
9996	9997	Rechargeable Lamp Scanner Vacuum	Appear country enter there pull thing computer.	Knapp-Key	Cycling	300.0	INR	177.0	4.707960e+12	MintCream	limited_st
9997	9998	Heater	Owner plan approach soldier church little little	Meyer, Case and Mills	Fitness Equipment	399.0	INR	996.0	5.286510e+12	LightGray	pre_or

```
# checking if there space
```

```
df['Brand'].unique().tolist()
```

```
'Saunders, Farley and Bradley',  
'Rosario-Velez',  
'Peters and Sons',  
'Abbott-Mccall',  
'Walker, Pittman and Wood',  
'Waller Ltd',  
'Golden, Booker and Green',  
'Norman, Dalton and Wolf',  
'Richardson, Knox and Parsons',  
'Hutchinson Ltd',  
'Mcmillan, Benton and Schneider',  
'Gilmore, Frey and Barron',  
'Carpenter Group',  
'Brady-Meyers',  
'Sanford, Pennington and Lang',  
'Barrera-Phelps',  
'Wu, McClure and Peters',  
'Mcdaniel-Shelton',  
'Weiss LLC',  
'Finley, Frey and Adkins',  
'Benton Inc',  
'Blevins, Meadows and Porter',  
'Pham-Robinson',  
'Archer, Bryant and Parker',  
'Lawson Group',  
'Norton LLC',  
'Randolph-Richardson',  
'Dixon-Pearson',  
'Anderson, Barker and Blevins',  
'Molina Group',  
'Drake-Stanley',  
'Simon, Andrade and Riggs',  
'West LLC',  
'Hernandez-Castro',  
'Villanueva-Steele',  
'Potter LLC',  
'Hobbs-Key',  
'Cruz-Sanford',  
'Robertson-Brown',  
'Huffman PLC',  
'Villegas-Stephenson',  
'Patton, Merritt and Baker',  
'Arnold-Melton',  
'Howe Group',  
'Hurst, Travis and Cervantes',  
'Frank and Sons',  
'Simpson LLC',  
'Ballard-Potter',  
'Keith-Howe',  
'Bradshaw LLC',  
'Brady PLC',  
'Goodwin Inc',  
'Wilkins PLC',  
'Frazier, Gamble and Meza',  
'Mcguire, Peterson and Pratt',  
'Fernandez-Russell',  
'Mann Ltd',  
'Hudson, Mueller and Vargas',
```

```
df['Category'].unique().tolist()
```

```
["Kids' Clothing",  
'Grooming Tools',  
'Fishing & Hunting',  
'Skincare',  
'Laptops & Computers',  
'Kitchen Appliances',  
'Clothing & Apparel',  
'Beauty & Personal Care',  
'Makeup',  
'Toys & Games',  
"Men's Clothing",  
'Camping & Hiking',  
'Cleaning Supplies',  
'Headphones & Earbuds',  
'Home & Kitchen',  
'Haircare',  
'Cameras & Accessories',  
'Bedding & Bath',  
'Accessories (Bags, Hats, Belts)',  
'Team Sports',  
'Smartwatches',  
'Fragrances',  
"Women's Clothing",  
'Shoes & Footwear',
```

```
'Furniture',
'Home Decor',
'Fitness Equipment',
'Health & Wellness',
'Smartphones',
'Office Supplies',
'Cycling',
'Automotive',
'Sports & Outdoors',
'Books & Stationery']
```

```
df['Brand'].value_counts()
```

count	
Brand	
Kennedy LLC	5
Buckley Inc	4
Watkins Group	4
Flores LLC	4
Johnston Ltd	4
...	...
Hendricks LLC	1
Mcconnell, Rangel and Brennan	1
Wall-Mack	1
Mcfarland-Henry	1
Graham Ltd	1

9241 rows × 1 columns

dtype: int64

```
df['Availability'].unique().tolist()
```

```
['limited_stock',
'discontinued',
'pre_order',
'out_of_stock',
'in_stock',
'backorder']
```

```
df['Availability'].value_counts()
```

count	
Availability	
discontinued	1706
out_of_stock	1700
pre_order	1673
limited_stock	1644
in_stock	1644
backorder	1633

dtype: int64

```
df['Color'].unique().tolist()
```

```
['Cornsilk',
'Bisque',
'Blue',
'Ivory',
'FloralWhite',
'LightSalmon',
'MediumPurple',
'PaleGreen',
'DarkBlue',
'SandyBrown',
'Thistle',
'OldLace',
'LightSkyBlue',
'MediumBlue',
'DarkTurquoise',
```



```
'Azure',
'DeepSkyBlue',
'DarkSeaGreen',
'BlanchedAlmond',
'LavenderBlush',
'LemonChiffon',
'SaddleBrown',
'CadetBlue',
'DarkSalmon',
'HoneyDew',
'DarkRed',
'Peru',
'AntiqueWhite',
'DarkCyan',
'LightGray',
'RoyalBlue',
'CornflowerBlue',
'Red',
'PaleGoldenRod',
'IndianRed',
'LimeGreen',
'Aqua',
'Snow',
'SpringGreen',
'Teal',
'DarkOrchid',
'LightGreen',
'GreenYellow',
'Plum',
'MediumOrchid',
'Fuchsia',
'MediumSpringGreen',
'LightBlue',
'Orchid',
'MediumTurquoise',
'RosyBrown',
'Crimson',
'Tomato',
'LightGoldenRodYellow',
'LightSeaGreen',
'SlateBlue',
'PowderBlue',
'Salmon'.
```

```
df['Color'].value_counts()
```

count	
Color	
Cornsilk	101
Peru	93
AntiqueWhite	92
Ivory	92
Salmon	88
...	...
MediumPurple	56
FloralWhite	55
Wheat	53
White	49
Turquoise	48

140 rows × 1 columns

dtype: int64

```
df['Name'].unique().tolist()
```

```
['Smart Fan Iron Cooker Go Wireless Portable',
'Fan',
'Smart Speakerphone Charger Eco Plus Clean',
'Premium Grill Trimmer Portable',
'Keyboard Freezer',
'Smart Fridge Plus',
'Compact Dock One Portable Wireless',
'Premium Cooler Digital Portable X',
'Compact Fan Scooter Headphones',
'Eco Dock Air',
'Thermostat Stove Lamp',
'Mini Powerbank Lite Plus',
'Scooter Bicycle Oven',
```

```
'Wireless Light Heater Clock X',
'Ultra Powerbank Scale Automatic',
'Projector',
'Blender',
'Wireless Speakerphone Oven Advanced',
'Iron Microphone',
'Lock',
'Automatic Speaker Wireless Plus Edge',
'Eco Treadmill Plus Smart',
'Smart Microphone Clock Ultra One',
'Digital Router Scanner One',
'Pro Stove Light',
'Clean Mixer Ultra Eco',
'Ultra Fan',
'Ultra Drone Smart',
'Smart Lock Mini Pro',
'Rechargeable Keyboard Toaster Monitor Portable Fast',
'Mini Mixer Freezer Trimmer Smart',
'Heater Cooker',
'Fast Shaver X One',
'Wireless Webcam Bicycle Camera',
'Clean Scale Speaker Powerbank Clean Wireless',
'Advanced Fan Powerbank',
'Pro Thermostat Speaker Touch Eco',
'Fast Blender Treadmill Tablet One Air',
'Automatic Iron Toaster Projector',
'Smart Scale Stove',
'Smart Webcam Mouse Digital',
'Compact Drone Lite',
'Digital Scale Wireless One Portable',
'Clean Drone Thermostat Headphones',
'Wireless Scooter',
'Clean Light Advanced Smart Premium',
'Compact Fan Iron Clock Digital',
'Ultra Watch Eco',
'Fast Lamp Scanner Vacuum',
'Eco Stove Edge',
'Fast Powerbank Radio Automatic Touch Ultra',
'Pro Oven Scale',
'Smart Printer Charger Brush Smart Ultra Touch',
'Wireless Watch Premium Air',
'Advanced Thermostat Microphone Scooter Clean Smart',
'Pro Mouse Mini Rechargeable',
'Silent Powerbank Radio Smart Plus',
'Clean Speakerphone Edge Rechargeable Max'.
```

```
df['Name'].value_counts()
```

	count
Name	
Fan	44
Dock	43
Freezer	34
Lock	32
Brush	27
...	...
Eco Grill Grill Portable	1
Mini Printer Lock Thermostat	1
Smart Grill Microphone Projector 360	1
Fast Powerbank Ultra	1
Smart Microphone X Portable Fast	1

7896 rows × 1 columns

dtype: int64

```
df['Description'].value_counts()
```

	count
Between key father half.	1
Part perform result newspaper.	1
Agreement ten responsibility whole maybe pull.	1
Who source better.	1
Expect public dinner.	1
...	...
Remain Congress blood plan voice.	1
Lawyer one than fire.	1
Quickly inside pull line lay start.	1
All movement yeah tax me.	1
Catch enough role nearly.	1

10000 rows × 1 columns

dtype: int64

df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 14 columns):
#   Column                Non-Null Count  Dtype  
---  -
0    Index                 10000 non-null  int64  
1    Name                  10000 non-null  object  
2    Description           10000 non-null  object  
3    Brand                 10000 non-null  object  
4    Category              10000 non-null  object  
5    MRP_Price             10000 non-null  float64 
6    Currency              10000 non-null  object  
7    Stock                 10000 non-null  float64 
8    EAN                   10000 non-null  float64 
9    Color                 10000 non-null  object  
10   Availability           10000 non-null  object  
11   Internal ID           10000 non-null  int64  
12   Qty                   10000 non-null  float64 
13   Selling_price         10000 non-null  float64 
dtypes: float64(5), int64(2), object(7)
memory usage: 1.1+ MB
```

change datatype

```
df['Stock'] = pd.to_numeric(df['Stock'], errors='coerce').astype('Int64')
df['Qty'] = pd.to_numeric(df['Qty'], errors='coerce').astype('Int64')
```

```
cat_cols = ['Brand', 'Category', 'Color', 'Availability', 'Currency']
```

```
for col in cat_cols:
    df[col] = df[col].astype('category')
```

```
df['Name'] = df['Name'].astype(str)
df['Description'] = df['Description'].astype(str)
```

```
df['MRP_Price'] = df['MRP_Price'].astype(float)
```

```
df['EAN'] = df['EAN'].astype('Int64').astype(str)
```

df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 14 columns):
#   Column                Non-Null Count  Dtype  
---  -
0    Index                 10000 non-null  int64  
1    Name                  10000 non-null  object  
2    Description           10000 non-null  object  
3    Brand                 10000 non-null  category
```

```
4 Category      10000 non-null category
5 MRP_Price     10000 non-null float64
6 Currency      10000 non-null category
7 Stock         10000 non-null Int64
8 EAN           10000 non-null object
9 Color         10000 non-null category
10 Availability  10000 non-null category
11 Internal ID   10000 non-null int64
12 Qty          10000 non-null Int64
13 Selling_price 10000 non-null float64
dtypes: Int64(2), category(5), float64(2), int64(2), object(3)
memory usage: 1.1+ MB
```

```
# column for category section
```

```
menu = {
    "Men's Clothing": "Apparel & Fashion",
    "Women's Clothing": "Apparel & Fashion",
    "Kids' Clothing": "Apparel & Fashion",
    "Shoes & Footwear": "Apparel & Fashion",
    "Clothing & Apparel": "Apparel & Fashion",
    "Accessories (Bags, Hats, Belts)": "Apparel & Fashion",

    "Beauty & Personal Care": "Beauty & Personal Care",
    "Skincare": "Beauty & Personal Care",
    "Makeup": "Beauty & Personal Care",
    "Haircare": "Beauty & Personal Care",
    "Fragrances": "Beauty & Personal Care",
    "Grooming Tools": "Beauty & Personal Care",

    "Home & Kitchen": "Home & Living",
    "Kitchen Appliances": "Home & Living",
    "Furniture": "Home & Living",
    "Home Decor": "Home & Living",
    "Bedding & Bath": "Home & Living",
    "Cleaning Supplies": "Home & Living",

    "Smartphones": "Mobiles & Electronics",
    "Laptops & Computers": "Mobiles & Electronics",
    "Smartwatches": "Mobiles & Electronics",
    "Headphones & Earbuds": "Mobiles & Electronics",
    "Cameras & Accessories": "Mobiles & Electronics",

    "Sports & Outdoors": "Sports, Fitness & Outdoors",
    "Fitness Equipment": "Sports, Fitness & Outdoors",
    "Cycling": "Sports, Fitness & Outdoors",
    "Camping & Hiking": "Sports, Fitness & Outdoors",
    "Fishing & Hunting": "Sports, Fitness & Outdoors",
    "Team Sports": "Sports, Fitness & Outdoors",

    "Toys & Games": "Toys, Books & Office",
    "Books & Stationery": "Toys, Books & Office",
    "Office Supplies": "Toys, Books & Office",

    "Automotive": "Automotive",
    "Health & Wellness": "Health & Wellness"
}

df["Category_Section"] = df["Category"].map(menu)
```

```
# column for availability
```

```
availability_map = {
    "in_stock": "Available",
    "limited_stock": "Available (Limited)",
    "pre_order": "Temporarily Unavailable",
    "backorder": "Temporarily Unavailable",
    "out_of_stock": "Unavailable",
    "discontinued": "Unavailable"
}

df["Availability_Status"] = df["Availability"].map(availability_map)
```

```
# df.sort_values('Availability_Status').head()
```

Show hidden output

```
df.info()
```

Show hidden output

```
# calculating discount
```

```
df['discount'] = df['MRP_Price'] - df['Selling_price']  
  
df[['MRP_Price', 'Selling_price', 'discount']].describe()
```

	MRP_Price	Selling_price	discount
count	10000.000000	10000.000000	10000.000000
mean	503.328300	365.351640	137.976660
std	289.702596	224.066094	111.648617
min	1.000000	0.500000	-309.400000
25%	251.000000	177.120000	47.600000
50%	506.000000	353.400000	107.470000
75%	756.000000	529.540000	204.410000
max	999.000000	949.050000	609.600000

```
# renaming column name
```

```
df.rename(columns={'Selling_price': 'Selling Price'}, inplace=True)  
df.rename(columns={'MRP_Price': 'MRP Price'}, inplace=True)  
df.rename(columns={'Category_Section': 'Category Section'}, inplace=True)  
df.rename(columns={'Availability_Status': 'Availability Status'}, inplace=True)
```

```
# save clean dataset
```

```
df.to_csv('Products(clean dataset).csv', index=False)
```

```
df.shape
```

Show hidden output

```
df.describe(include='all').T
```

Show hidden output

```
df.duplicated().sum()
```

Show hidden output

```
#Analysing and providing the stats of each column.
```

```
#stats for Name column  
name_stats = {  
    'unique_values': df['Name'].nunique(),  
    'missing_values': df['Name'].isnull().sum(),  
    'avg_length': df['Name'].dropna().str.len().mean(),  
    'min_length': df['Name'].dropna().str.len().min(),  
    'max_length': df['Name'].dropna().str.len().max(),  
    'most_frequent': df['Name'].mode().iloc[0]  
}  
  
name_stats
```

```
{'unique_values': 7896,  
 'missing_values': np.int64(0),  
 'avg_length': np.float64(26.1114),  
 'min_length': 3,  
 'max_length': 64,  
 'most_frequent': 'Fan'}
```

```
name_stats_df = pd.DataFrame([name_stats], index=['Name'])
name_stats_df
```

	unique_values	missing_values	avg_length	min_length	max_length	most_frequent
Name	7896	0	26.1114	3	64	Fan

```
top_5_frequent= df['Name'].value_counts().head(5).to_dict()
top_5_frequent
```

```
{'Fan': 44, 'Dock': 43, 'Freezer': 34, 'Lock': 32, 'Brush': 27}
```

```
#stats for Description column
description_stats = {
    'unique_values': df['Description'].nunique(),
    'missing_values': df['Description'].isnull().sum(),
    'avg_length': df['Description'].dropna().str.len().mean(),
    'min_length': df['Description'].dropna().str.len().min(),
    'max_length': df['Description'].dropna().str.len().max(),
    'most_frequent': df['Description'].mode().iloc[0]
}

description_stats
```

```
{'unique_values': 10000,
 'missing_values': np.int64(0),
 'avg_length': np.float64(35.9708),
 'min_length': 11,
 'max_length': 78,
 'most_frequent': 'A beyond become company blue.'}
```

```
description_stats_stats_df = pd.DataFrame([description_stats], index=['Description'])
description_stats_stats_df
```

	unique_values	missing_values	avg_length	min_length	max_length	most_frequent
Description	10000	0	35.9708	11	78	A beyond become company blue.

```
#stats for Brand column
brand_stats = {
    'unique_values': df['Brand'].nunique(),
    'missing_values': df['Brand'].isnull().sum(),
    'avg_length': df['Brand'].dropna().str.len().mean(),
    'min_length': df['Brand'].dropna().str.len().min(),
    'max_length': df['Brand'].dropna().str.len().max(),
    'most_frequent': df['Brand'].mode().iloc[0]
}

brand_stats
```

```
{'unique_values': 9241,
 'missing_values': np.int64(0),
 'avg_length': np.float64(16.5369),
 'min_length': 5,
 'max_length': 35,
 'most_frequent': 'Kennedy LLC'}
```

```
brand_stats_df = pd.DataFrame([brand_stats], index=['Brand'])
brand_stats_df
```

	unique_values	missing_values	avg_length	min_length	max_length	most_frequent
Brand	9241	0	16.5369	5	35	Kennedy LLC

```
#stats for Category column

category_stats = {
    'unique_values': df['Category'].nunique(),
    'missing_values': df['Category'].isnull().sum(),
    'avg_length': df['Category'].dropna().str.len().mean(),
    'min_length': df['Category'].dropna().str.len().min(),
    'max_length': df['Category'].dropna().str.len().max(),
    'most_frequent': df['Category'].mode().iloc[0]
}
```

```
category_stats
```

```
{'unique_values': 34,
 'missing_values': np.int64(0),
 'avg_length': np.float64(14.6677),
 'min_length': 6,
 'max_length': 31,
 'most_frequent': 'Clothing & Apparel'}
```

```
category_stats_df = pd.DataFrame([category_stats], index=['Category'])
category_stats_df
```

	unique_values	missing_values	avg_length	min_length	max_length	most_frequent
Category	34	0	14.6677	6	31	Clothing & Apparel

```
# Top categories
df['Category'].value_counts().head(5)
```

	count
Category	
Clothing & Apparel	320
Kitchen Appliances	319
Home & Kitchen	316
Team Sports	316
Grooming Tools	311

```
dtype: int64
```

```
#stats for MRP_Price column
mrp_stats = {
    'unique_values': df['MRP_Price'].nunique(),
    'missing_values': df['MRP_Price'].isnull().sum(),
    'mean': df['MRP_Price'].mean(),
    'median': df['MRP_Price'].median(),
    'min': df['MRP_Price'].min(),
    'max': df['MRP_Price'].max(),
    'std_dev': df['MRP_Price'].std(),
    'skewness': df['MRP_Price'].skew()
}
```

```
mrp_stats

{'unique_values': 999,
 'missing_values': np.int64(0),
 'mean': np.float64(503.3283),
 'median': 506.0,
 'min': 1.0,
 'max': 999.0,
 'std_dev': 289.702595567417,
 'skewness': np.float64(-0.016483312382078688)}
```

```
num_stats_df = pd.DataFrame([mrp_stats], index=['MRP_Price'])
num_stats_df
```

	unique_values	missing_values	mean	median	min	max	std_dev	skewness
MRP_Price	999	0	503.3283	506.0	1.0	999.0	289.702596	-0.016483

```
# stats for Stock column
stock_stats = {
    'unique_values': df['Stock'].nunique(),
    'missing_values': df['Stock'].isnull().sum(),
    'mean': df['Stock'].mean(),
    'median': df['Stock'].median(),
    'min': df['Stock'].min(),
    'max': df['Stock'].max(),
    'std_dev': df['Stock'].std(),
    'zero_stock_count': (df['Stock'] == 0).sum()
}
```

```
stock_stats
```

```
{'unique_values': 1000,
 'missing_values': np.int64(0),
```

```
'mean': np.float64(498.0426),
'median': np.float64(499.0),
'min': np.int64(0),
'max': np.int64(999),
'std_dev': np.float64(288.65579329356996),
'zero_stock_count': np.int64(22)}
```

```
stock_stats_df = pd.DataFrame([stock_stats], index=['Stock'])
stock_stats_df
```

	unique_values	missing_values	mean	median	min	max	std_dev	zero_stock_count
Stock	1000	0	498.0426	499.0	0	999	288.655793	22

```
# stats for EAN column
ean_stats = {
    'unique_values': df['EAN'].nunique(),
    'missing_values': df['EAN'].isnull().sum(),
    'duplicate_rows': df['EAN'].duplicated(keep=False).sum()
}

ean_stats
```

```
{'unique_values': 9952,
 'missing_values': np.int64(0),
 'duplicate_rows': np.int64(96)}
```

```
ean_stats_df = pd.DataFrame([ean_stats], index=['EAN'])
ean_stats_df
```

	unique_values	missing_values	duplicate_rows
EAN	9952	0	96

```
#stats for color column
color_stats = {
    'unique_values': df['Color'].nunique(),
    'missing_values': df['Color'].isnull().sum(),
    'avg_length': df['Color'].dropna().str.len().mean(),
    'min_length': df['Color'].dropna().str.len().min(),
    'max_length': df['Color'].dropna().str.len().max(),
    'most_frequent': df['Color'].mode().iloc[0]
}

color_stats
```

```
{'unique_values': 140,
 'missing_values': np.int64(0),
 'avg_length': np.float64(8.9051),
 'min_length': 3,
 'max_length': 20,
 'most_frequent': 'Cornsilk'}
```

```
color_stats_df = pd.DataFrame([color_stats], index=['Color'])
color_stats_df
```

	unique_values	missing_values	avg_length	min_length	max_length	most_frequent
Color	140	0	8.9051	3	20	Cornsilk

```
# Check inconsistent naming
df['Color'].dropna().str.lower().value_counts().head(10)
```


count	
Color	
<hr/>	
cornsilk	101
peru	93
antiquewhite	92
ivory	92
salmon	88
plum	88
darkslategray	87
maroon	85
teal	85
lightcoral	85

dtype: int64

```
#stats for Availability column
availability_stats = {
    'unique_values': df['Availability'].nunique(),
    'missing_values': df['Availability'].isnull().sum(),
    'avg_length': df['Availability'].dropna().str.len().mean(),
    'min_length': df['Availability'].dropna().str.len().min(),
    'max_length': df['Availability'].dropna().str.len().max(),
    'most_frequent': df['Availability'].mode().iloc[0]
}
```

availability_stats

```
{'unique_values': 6,
 'missing_values': np.int64(0),
 'avg_length': np.float64(10.515),
 'min_length': 8,
 'max_length': 13,
 'most_frequent': 'discontinued'}
```

```
availability_stats_df = pd.DataFrame([availability_stats], index=['Availability'])
availability_stats_df
```

	unique_values	missing_values	avg_length	min_length	max_length	most_frequent
Availability	6	0	10.515	8	13	discontinued

```
#stats for Internal ID
internal_id_stats = {
    'unique_values': df['Internal ID'].nunique(),
    'missing_values': df['Internal ID'].isnull().sum(),
    'duplicate_rows': df['Internal ID'].duplicated(keep=False).sum(),
}
```

internal_id_stats

```
{'unique_values': 99,
 'missing_values': np.int64(0),
 'duplicate_rows': np.int64(10000)}
```

```
internal_id_stats_df = pd.DataFrame([internal_id_stats], index=['Internal ID'])
internal_id_stats_df
```

	unique_values	missing_values	duplicate_rows
Internal ID	99	0	10000

```
# stats for Qty column
qty_stats = {
    'unique_values': df['Qty'].nunique(),
    'missing_values': df['Qty'].isnull().sum(),
    'mean': df['Qty'].mean(),
    'median': df['Qty'].median(),
    'min': df['Qty'].min(),
    'max': df['Qty'].max(),
}
```

```

        'std_dev': df['Qty'].std(),
        'skewness': df['Qty'].skew()
    }

```

qty_stats

```

{'unique_values': 952,
 'missing_values': np.int64(0),
 'mean': np.float64(523.2285),
 'median': np.float64(522.0),
 'min': np.int64(0),
 'max': np.int64(1000),
 'std_dev': np.float64(276.74841509261734),
 'skewness': np.float64(-0.0014358144305663615)}

```

```

qty_stats_df = pd.DataFrame([qty_stats], index=['Qty'])
qty_stats_df

```

	unique_values	missing_values	mean	median	min	max	std_dev	skewness
Qty	952	0	523.2285	522.0	0	1000	276.748415	-0.001436

```

# stats for Selling_Price column
selling_price_stats = {
    'unique_values': df['Selling_price'].nunique(),
    'missing_values': df['Selling_price'].isnull().sum(),
    'mean': df['Selling_price'].mean(),
    'median': df['Selling_price'].median(),
    'min': df['Selling_price'].min(),
    'max': df['Selling_price'].max(),
    'std_dev': df['Selling_price'].std()
}
selling_price_stats

```

```

{'unique_values': 7915,
 'missing_values': np.int64(0),
 'mean': np.float64(365.69653999999997),
 'median': 353.94500000000005,
 'min': 0.5,
 'max': 963.0,
 'std_dev': 224.39278000786976}

```

```

selling_price_stats_df = pd.DataFrame([selling_price_stats], index=['Selling_price'])
selling_price_stats_df

```

	unique_values	missing_values	mean	median	min	max	std_dev
Selling_price	7915	0	365.69654	353.945	0.5	963.0	224.39278

```

# stats for column Category_Section
category_section_stats = {
    'unique_values': df['Category_Section'].nunique(),
    'missing_values': df['Category_Section'].isnull().sum(),
    'avg_length': df['Category_Section'].astype(str).str.len().mean(),
    'min_length': df['Category_Section'].astype(str).str.len().min(),
    'max_length': df['Category_Section'].astype(str).str.len().max(),
    'most_frequent': df['Category_Section'].mode().iloc[0]
}
category_section_stats

```

```

{'unique_values': 8,
 'missing_values': np.int64(0),
 'avg_length': np.float64(19.3901),
 'min_length': 10,
 'max_length': 26,
 'most_frequent': 'Home & Living'}

```

```

category_section_stats_df = pd.DataFrame([category_section_stats], index=['Category_Section'])
category_section_stats_df

```

Show hidden output

```

# stats for column Availability_Status
availability_status_stats = {
    'unique_values': df['Availability_Status'].nunique(),
    'missing_values': df['Availability_Status'].isnull().sum(),
    'avg_length': df['Availability_Status'].astype(str).str.len().mean(),
    'min_length': df['Availability_Status'].astype(str).str.len().min(),
    'max_length': df['Availability_Status'].astype(str).str.len().max(),

```

```
'most_frequent': df['Availability_Status'].mode().iloc[0]
}
```

availability_status_stats

```
{'unique_values': 4,
 'missing_values': np.int64(0),
 'avg_length': np.float64(15.9536),
 'min_length': 9,
 'max_length': 23,
 'most_frequent': 'Unavailable'}
```

```
availability_status_stats_df = pd.DataFrame([availability_status_stats], index=['Availability_Status'])
availability_status_stats_df
```

[Show hidden output](#)

```
(dff['Selling price'] > dff['MRP Price']).sum()
```