

# Data Cleaning Workflow Template

This notebook is a reusable template for cleaning raw data:

1. Load data
2. Inspect structure
3. Handle missing values
4. Remove duplicates
5. Fix data types
6. Clean text/categories
7. Detect & cap outliers (IQR)
8. Validate & export

```
In [11]: # === 1. IMPORTS & SETTINGS ===  
  
import pandas as pd  
import numpy as np  
import matplotlib.pyplot as plt  
import seaborn as sns  
  
pd.set_option('display.max_columns', None)  
pd.set_option('display.float_format', lambda x: f'{x:.2f}')  
  
print("Libraries loaded.")
```


Libraries loaded.

```
In [6]: 3# === 2. LOAD RAW DATA ===  
# TODO: Change the file name/path for each new project.  
  
DATA_PATH = "/home/jkatz015/repos/personal/portfolio/projects/project-02-ap-automation/data/invoices_raw/ap_transactions.csv"  
df = pd.read_csv(DATA_PATH)  
  
print("Rows:", len(df))  
df.head()
```

Rows: 50150

Out[6]:

	INVOICE_ID	VENDOR_ID	VENDOR_NAME	VENDOR_NUMBER	VENDOR_COUNTRY	VENDOR_REGION	INVOICE_DATE	DUE_DATE
0	873641333	846210656	Metro Solutions	1230020547	Brunei Darussalam	North-Region	2022-04-11	2022-05-11
1	865186916	892667166	Metro Corp	1230022657	Spain	North-Region	2022-03-09	2022-04-09
2	914629138	921212744	Superior Enterprises	1230002613	Kenya	South-Region	2022-02-23	2022-03-23
3	861070279	803387626	Global Manufacturing	1230014342	Turkmenistan	North-Region	2022-02-01	2022-03-01
4	784182066	942636818	Quality Partners	1230014342	Guam	North-Region	2022-02-14	2022-03-14



In [12]: # === 3. BASIC STRUCTURE CHECK ===

```
df.info()
df.describe()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 50150 entries, 0 to 50149
Data columns (total 20 columns):
#   Column                Non-Null Count  Dtype
---  -
0   INVOICE_ID            50150 non-null  int64
1   VENDOR_ID             50150 non-null  int64
2   VENDOR_NAME           50150 non-null  object
3   VENDOR_NUMBER         50150 non-null  int64
4   VENDOR_COUNTRY        50150 non-null  object
5   VENDOR_REGION         50150 non-null  object
6   INVOICE_DATE          50150 non-null  object
7   DUE_DATE              50150 non-null  object
8   INVOICE_AMOUNT        50150 non-null  float64
9   CURRENCY              50150 non-null  object
10  GL_CODE               50150 non-null  object
11  PO_TYPE               50150 non-null  object
12  COMPANY_CODE          50150 non-null  int64
13  BUSINESS_UNIT         50150 non-null  int64
14  COST_CENTER           50150 non-null  object
15  PAYMENT_STATUS        50150 non-null  object
16  APPROVAL_STATUS       50150 non-null  float64
17  CREDIT_LIMIT          50150 non-null  float64
18  INVOICE_TIME          50150 non-null  int64
19  _ANOMALY_TYPE         50150 non-null  object
dtypes: float64(3), int64(6), object(11)
memory usage: 7.7+ MB
```

Out[12]:

	INVOICE_ID	VENDOR_ID	VENDOR_NUMBER	INVOICE_AMOUNT	COMPANY_CODE	BUSINESS_UNIT	APPROVAL_STATI
<b>count</b>	50150.00	50150.00	50150.00	50150.00	50150.00	50150.00	50150.
<b>mean</b>	879640407.81	876513993.64	1111686783.56	6317.53	3415.36	3304.16	66.
<b>std</b>	86465713.92	70885396.17	357815061.23	216296.24	594.99	696.53	8.
<b>min</b>	753460934.00	753458315.00	12100061.00	0.01	59.00	2100.00	52.
<b>25%</b>	815645178.75	815235731.75	1230011793.00	189.04	3260.00	2702.00	64.
<b>50%</b>	877306681.50	876551085.50	1230014314.00	457.02	3660.00	3300.00	64.
<b>75%</b>	938984852.25	938060359.25	1230014342.00	1057.20	3660.00	3908.00	64.
<b>max</b>	1897434115.00	999007452.00	1230025103.00	44854887.00	4260.00	4510.00	93.



In [8]:

```
# === 4. MISSING VALUES OVERVIEW ===

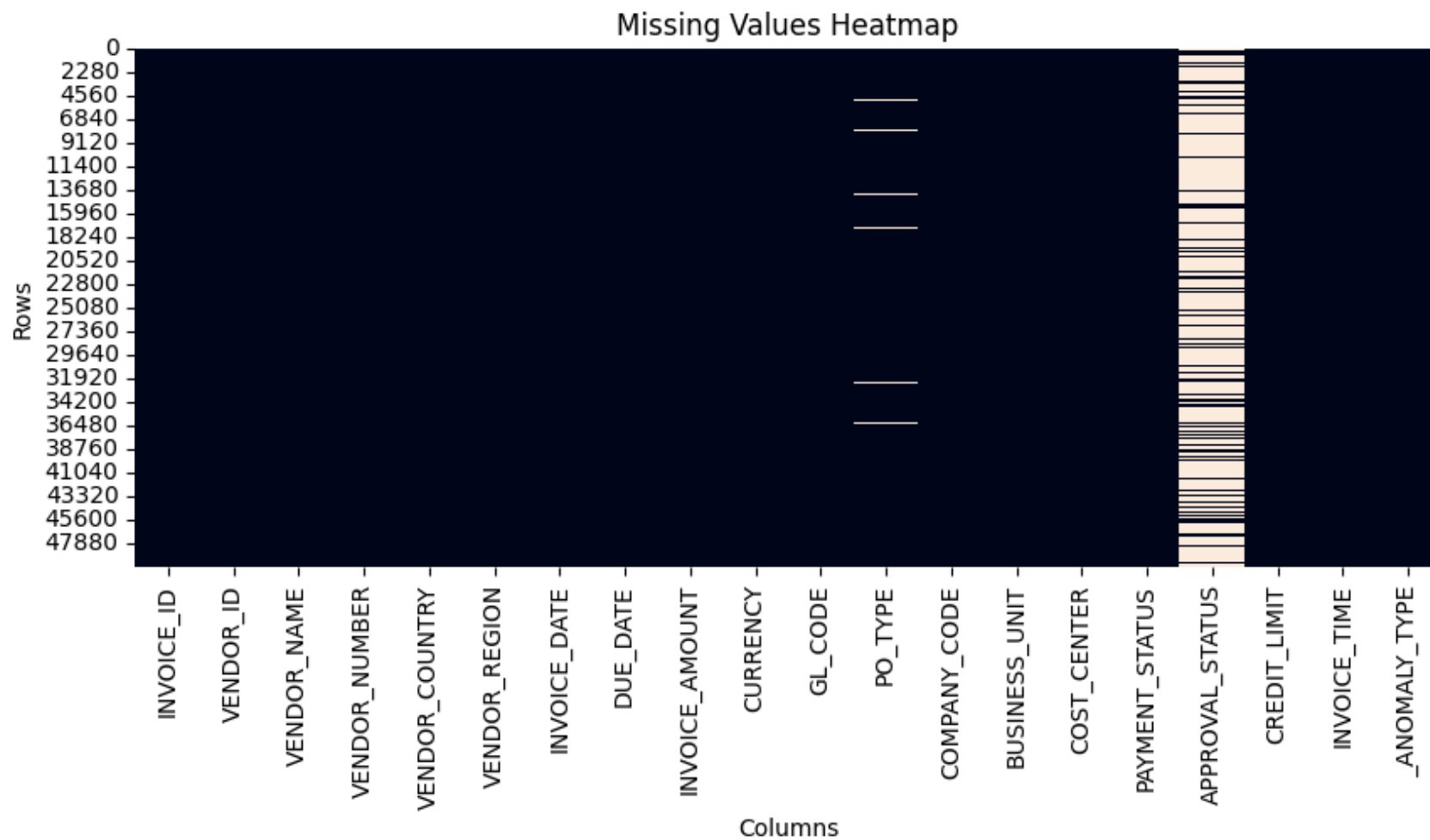
print("Missing values per column:")
print(df.isnull().sum())

plt.figure(figsize=(10, 4))
sns.heatmap(df.isnull(), cbar=False)
plt.title("Missing Values Heatmap")
plt.xlabel("Columns")
plt.ylabel("Rows")
plt.show()
```

Missing values per column:

INVOICE_ID	0
VENDOR_ID	0
VENDOR_NAME	0
VENDOR_NUMBER	0
VENDOR_COUNTRY	0
VENDOR_REGION	0
INVOICE_DATE	0
DUE_DATE	0
INVOICE_AMOUNT	0
CURRENCY	0
GL_CODE	0
PO_TYPE	1063
COMPANY_CODE	0
BUSINESS_UNIT	0
COST_CENTER	0
PAYMENT_STATUS	0
APPROVAL_STATUS	38811
CREDIT_LIMIT	0
INVOICE_TIME	0
_ANOMALY_TYPE	0

dtype: int64



In [13]: # === 5. IDENTIFY COLUMN TYPES ===

```
numeric_cols = df.select_dtypes(include=['float64', 'int64', 'Int64']).columns.tolist()
text_cols     = df.select_dtypes(include=['object']).columns.tolist()

print("Numeric columns:", numeric_cols)
print("Text columns:", text_cols)
```

Numeric columns: ['INVOICE\_ID', 'VENDOR\_ID', 'VENDOR\_NUMBER', 'INVOICE\_AMOUNT', 'COMPANY\_CODE', 'BUSINESS\_UNIT', 'APPROVAL\_STATUS', 'CREDIT\_LIMIT', 'INVOICE\_TIME']  
Text columns: ['VENDOR\_NAME', 'VENDOR\_COUNTRY', 'VENDOR\_REGION', 'INVOICE\_DATE', 'DUE\_DATE', 'CURRENCY', 'GL\_CODE', 'PO\_TYPE', 'COST\_CENTER', 'PAYMENT\_STATUS', '\_ANOMALY\_TYPE']

```
In [14]: # === 6. HANDLE MISSING VALUES ===  
# Adjust per project if needed.  
  
# 6a. Numeric → fill with median  
for col in numeric_cols:  
    if df[col].isnull().sum() > 0:  
        median_value = df[col].median()  
        df.loc[:, col] = df[col].fillna(median_value)  
  
# 6b. Text → fill with "Unknown"  
for col in text_cols:  
    if df[col].isnull().sum() > 0:  
        df.loc[:, col] = df[col].fillna("Unknown")  
  
print("Missing values after fill:")  
print(df.isnull().sum())
```

Missing values after fill:

INVOICE_ID	0
VENDOR_ID	0
VENDOR_NAME	0
VENDOR_NUMBER	0
VENDOR_COUNTRY	0
VENDOR_REGION	0
INVOICE_DATE	0
DUE_DATE	0
INVOICE_AMOUNT	0
CURRENCY	0
GL_CODE	0
PO_TYPE	0
COMPANY_CODE	0
BUSINESS_UNIT	0
COST_CENTER	0
PAYMENT_STATUS	0
APPROVAL_STATUS	0
CREDIT_LIMIT	0
INVOICE_TIME	0
_ANOMALY_TYPE	0

dtype: int64

In [15]: *# === 7. REMOVE DUPLICATES ===*

```
dupes_before = df.duplicated().sum()
print("Duplicates before:", dupes_before)

df = df.drop_duplicates()

dupes_after = df.duplicated().sum()
print("Duplicates after:", dupes_after)
```

Duplicates before: 0

Duplicates after: 0

In [16]: *# === 8. FIX DATA TYPES ===*

```
# Example: convert columns whose name suggests date/time
date_like_cols = [c for c in df.columns if 'date' in c.lower()]
for col in date_like_cols:
    df.loc[:, col] = pd.to_datetime(df[col], errors='coerce')
```



```
# Example: convert typical integer-like columns if present
int_candidate_cols = ['age', 'quantity', 'count'] # edit per dataset
for col in int_candidate_cols:
    if col in df.columns:
        df.loc[:, col] = df[col].astype('Int64')

print(df.dtypes)
```

```
INVOICE_ID          int64
VENDOR_ID           int64
VENDOR_NAME         object
VENDOR_NUMBER       int64
VENDOR_COUNTRY      object
VENDOR_REGION       object
INVOICE_DATE        object
DUE_DATE            object
INVOICE_AMOUNT      float64
CURRENCY            object
GL_CODE             object
PO_TYPE             object
COMPANY_CODE        int64
BUSINESS_UNIT       int64
COST_CENTER         object
PAYMENT_STATUS      object
APPROVAL_STATUS     float64
CREDIT_LIMIT        float64
INVOICE_TIME        int64
_ANOMALY_TYPE       object
dtype: object
```

In [19]: # === 9. CLEAN TEXT / CATEGORICAL COLUMNS ===

```
for col in text_cols:
    df.loc[:, col] = df[col].astype(str)
    df.loc[:, col] = df[col].str.strip()
    df.loc[:, col] = df[col].str.upper()

# Example: inspect a key category, if it exists
if 'CITY' in df.columns:
    print(df['CITY'].value_counts().head(20))
```

```
In [20]: # === 10. OUTLIER DETECTION & CAPPING (IQR) ===

def iqr_bounds(series: pd.Series):
    Q1 = series.quantile(0.25)
    Q3 = series.quantile(0.75)
    IQR = Q3 - Q1
    lower = Q1 - 1.5 * IQR
    upper = Q3 + 1.5 * IQR
    return lower, upper

for col in numeric_cols:
    if df[col].nunique() <= 1:
        continue # skip constant columns

    lower, upper = iqr_bounds(df[col])
    print(f"{col}: lower={lower:.2f}, upper={upper:.2f}")

    df.loc[df[col] < lower, col] = lower
    df.loc[df[col] > upper, col] = upper
```

```
INVOICE_ID: lower=630635668.50, upper=1123994362.50
VENDOR_ID: lower=630998790.50, upper=1122297300.50
VENDOR_NUMBER: lower=1230007969.50, upper=1230018165.50
INVOICE_AMOUNT: lower=-1113.21, upper=2359.44
COMPANY_CODE: lower=2660.00, upper=4260.00
BUSINESS_UNIT: lower=893.00, upper=5717.00
INVOICE_TIME: lower=-67351.12, upper=337595.88
```

```
In [21]: # === 11. FINAL VALIDATION ===

print("Missing values per column:")
print(df.isnull().sum())

print("\nDuplicates:", df.duplicated().sum())

print("\nData types:")
print(df.dtypes)

print("\nPreview:")
df.head()
```

Missing values per column:

INVOICE_ID	0
VENDOR_ID	0
VENDOR_NAME	0
VENDOR_NUMBER	0
VENDOR_COUNTRY	0
VENDOR_REGION	0
INVOICE_DATE	0
DUE_DATE	0
INVOICE_AMOUNT	0
CURRENCY	0
GL_CODE	0
PO_TYPE	0
COMPANY_CODE	0
BUSINESS_UNIT	0
COST_CENTER	0
PAYMENT_STATUS	0
APPROVAL_STATUS	0
CREDIT_LIMIT	0
INVOICE_TIME	0
_ANOMALY_TYPE	0

dtype: int64

Duplicates: 0

Data types:

INVOICE_ID	float64
VENDOR_ID	float64
VENDOR_NAME	object
VENDOR_NUMBER	float64
VENDOR_COUNTRY	object
VENDOR_REGION	object
INVOICE_DATE	object
DUE_DATE	object
INVOICE_AMOUNT	float64
CURRENCY	object
GL_CODE	object
PO_TYPE	object
COMPANY_CODE	int64
BUSINESS_UNIT	int64
COST_CENTER	object
PAYMENT_STATUS	object

```


APPROVAL_STATUS    float64
CREDIT_LIMIT        float64
INVOICE_TIME        float64
_ANOMALY_TYPE       object
dtype: object

```

Preview:

Out[21]:

	INVOICE_ID	VENDOR_ID	VENDOR_NAME	VENDOR_NUMBER	VENDOR_COUNTRY	VENDOR_REGION	INVOICE_DATE	DUI
0	873641333.00	846210656.00	METRO SOLUTIONS	1230018165.50	BRUNEI DARUSSALAM	NORTH-REGION	2022-04-11 00:00:00	20
1	865186916.00	892667166.00	METRO CORP	1230018165.50	SPAIN	NORTH-REGION	2022-03-09 00:00:00	20
2	914629138.00	921212744.00	SUPERIOR ENTERPRISES	1230007969.50	KENYA	SOUTH-REGION	2022-02-23 00:00:00	20
3	861070279.00	803387626.00	GLOBAL MANUFACTURING	1230014342.00	TURKMENISTAN	NORTH-REGION	2022-02-01 00:00:00	20
4	784182066.00	942636818.00	QUALITY PARTNERS	1230014342.00	GUAM	NORTH-REGION	2022-02-14 00:00:00	20



In [ ]: