HR Data Cleaning Utilities (v 1 . 0)

This notebook demonstrates how to **simulate messy HR data** and then build a cleaning pipeline make it analysis-ready.

Data cleaning is a critical step in People Analytics — poor quality data = misleading insights.

```
In [1]: # Import libraries
   import pandas as pd
   import numpy as np
   import seaborn as sns
   import matplotlib.pyplot as plt
   import random
```

Create a Messy HR Dataset

We start from the processed dataset and intentionally add issues:

- Duplicates
- Missing values
- Inconsistent casing
- Invalid dates

Messy dataset created !

Explore the Messy Data

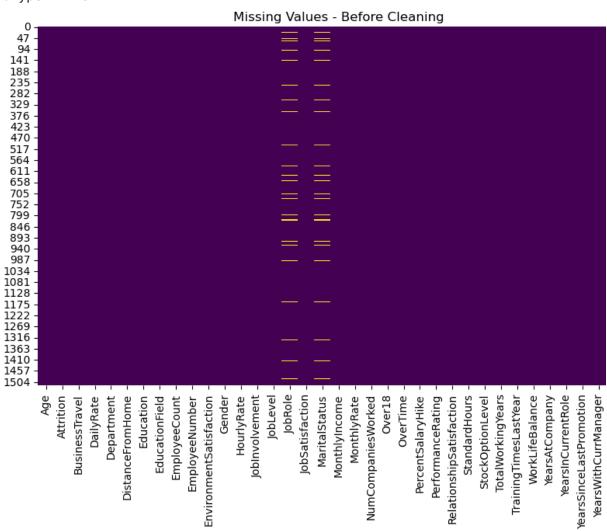
Before cleaning, let's check shape, null values, and visualize missing data.

```
In [6]: df = pd.read_csv("messy_hr_data.csv")
    print("Initial shape:", df.shape)
    print(df.isnull().sum().head())
```

```
plt.figure(figsize=(10,6))
sns.heatmap(df.isnull(), cbar=False, cmap="viridis")
plt.title("Missing Values - Before Cleaning")
plt.savefig("images/missing_values_before.png", dpi=300, bbox_inches="tig
plt.show()
```

Initial shape: (1520, 37) Age 0 Attrition BusinessTravel 0 0 DailyRate 0 Department

dtype: int64



Apply Cleaning Steps

Now we:

- 1. Remove duplicates
- 2 . Fill missing values
- 3 . Normalize categorical values
- 4 . Convert date columns into proper format

```
In [7]: # 1. Drop duplicates
        df = df.drop_duplicates()
        # 2. Fill missing values
```

```
df = df.fillna({
    "JobRole": "Unknown",
    "MaritalStatus": "Unknown"
})

# 3. Normalize categorical values
df["Department"] = df["Department"].str.title().str.strip()
df["JobRole"] = df["JobRole"].str.title().str.strip()

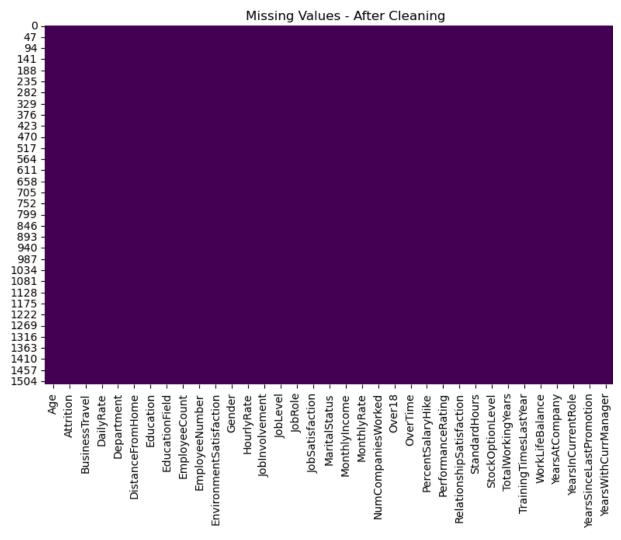
# 4. Fix Date column
df["DateOfJoining"] = pd.to_datetime(df["DateOfJoining"], errors="coerce")
```

Verify the Cleaning

Check if missing values reduced and visualize again.

```
In [8]: plt.figure(figsize=(10,6))
    sns.heatmap(df.isnull(), cbar=False, cmap="viridis")
    plt.title("Missing Values - After Cleaning")
    plt.savefig("images/missing_values_after.png", dpi=300, bbox_inches="tigh
    plt.show()

    print("Final shape:", df.shape)
```



Final shape: (1519, 37)

Export the Cleaned Dataset

The cleaned dataset can now be used for further People Analytics projects.

```
In [9]: df.to_csv("cleaned_hr_data.csv", index=False)
    print("Cleaning complete! Cleaned dataset saved as cleaned_hr_data.csv")
```

Cleaning complete! Cleaned dataset saved as cleaned_hr_data.csv

Create a Before -After Collage

```
In [10]: from PIL import Image

# Load images
before = Image.open("images/missing_values_before.png")
after = Image.open("images/missing_values_after.png")

# Resize to same height
h = min(before.height, after.height)
before = before.resize((int(before.width * h / before.height), h))
after = after.resize((int(after.width * h / after.height), h))

# Combine side by side
collage = Image.new("RGB", (before.width + after.width, h))
collage.paste(before, (0, 0))
collage.paste(after, (before.width, 0))
```

```
In [11]: # Save collage
    collage.save("images/missing_values_collage.png")
    print("Collage saved at images/missing_values_collage.png")
```

Collage saved at images/missing values collage.png



- Automated pipeline successfully cleaned the dataset.
- Issues fixed: duplicates, missing values, inconsistent casing, invalid dates.
- Before-After Collage created

```
In [ ]:
```