## IS-5960-04: MRP

# **Employability Analytics Project**

## **Data Validation & Transformation Documentation**

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#### **Tools Used:**

- Python
- Power BI Power Query Editor
- DAX Measures

## **Python Data Cleaning Steps**

Below are the Python scripts which we have executed for data cleaning.

## **Step 1: Load the Dataset**

import pandas as pd file path

=

"/Users/ananya/Documents/TEAM16\_MRP/Business\_analyst\_job\_listings\_linkedin.csv" df = pd.read\_csv(file\_path, dtype=str) **Output:** 

```
✓ Dataset Loaded Successfully!

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 921 entries, 0 to 920

Data columns (total 10 columns):
```

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#	Column	Non-Null Count	Dtype
0	title	921 non-null	object
1	location	921 non-null	object
2	publishedAt	921 non-null	object
3	companyName	911 non-null	object
4	description	921 non-null	object
5	applicationsCount	921 non-null	object
6	contractType	921 non-null	object
7	experienceLevel	921 non-null	object
8	workType	921 non-null	object
9	sector	921 non-null	object

**Step 2: Check for Missing Values Before** 

Cleaning: print("Missing Values:\n",

df.isnull().sum()) Output:

```
Missing Values:
 title
                        0
location
                       0
publishedAt
                       0
companyName
                      10
description
applicationsCount
contractType
                       0
experienceLevel
                       0
workType
sector
```

### **After Cleaning:**

df["companyName"] = df["companyName"].fillna("Unknown") print("Missing Values After Cleaning:\n", df.isnull().sum())

#### **Output:**

```
Missing Values After Cleaning:
 title
location
                      0
publishedAt
                      0
                      0
companyName
description
                      0
applicationsCount
                      0
contractType
                      0
                      0
experienceLevel
workType
                      0
sector
```

## Step 3: Convert applicationsCount to Numeric

```
import re def extract_number(val):
    match = re.search(r'\d+', str(val))         return
int(match.group()) if match else None
df["applicationsCount"] = df["applicationsCount"].apply(extract_number)
```

#### Step 4: Convert publishedAt to Standard Date Format

```
df["publishedAt"] = pd.to\_datetime(df["publishedAt"], format="\%m/\%d/\%y", errors="coerce")
```

## **Output:**

```
publishedAt
    2024-09-04
    2024-08-23
 1
 2
    2024-08-02
    2024-08-20
    2024-08-27
Step 5: Check and Fix Duplicate Job Listings Before fixing: duplicates =
df[df.duplicated(subset=["title", "location", "companyName"], keep=False)]
print("Duplicate Records Found:", len(duplicates))
Output:
 ... df.drop_duplicates(subset=["title", "location", "companyName"], keep="first", inplace=True)

▼ Duplicate Records Found: 201

After fixing: df.drop_duplicates(subset=["title", "location", "companyName"], keep="first",
inplace=True) duplicates = df[df.duplicated(subset=["title", "location", "companyName"],
keep=False)] print("Duplicate Records Found:", len(duplicates)) Output:
 0
      200
      200
 2 3 4
      170
      200
      200
      200
 8
      200
      200
 Name: applicationsCount, dtype: int64
 Duplicate Records Found: 0
Step 6: Save the Cleaned Dataset cleaned file path
```

"/Users/ananya/Documents/TEAM16\_MRP/CLEANED\_Business\_analyst\_job\_listings.csv" df.to\_csv(cleaned\_file\_path, index=False) print(f"Cleaned dataset saved at: {cleaned\_file\_path}")

#### **Transformations using Power BI (Power Query)**

After validation, we imported the cleaned CSV into Power BI and used Power Query to apply key transformations:

- Renamed columns for readability (e.g., publishedAt  $\rightarrow$  Published Date)
- Filtered out incomplete rows if any appeared during slicing
- Created calculated columns, such as:

SimulatedSalary using conditional logic based on experience level

We used "Transform Data" to create and clean dimensions, especially for visual compatibility across pages.

#### **DAX Measures in Power BI**

To support dynamic visuals, we created several DAX measures. Here are a few examples:

- Total Job Listings = COUNT('jobs'[title])
- Total Applications = SUM('jobs'[applicationsCount])
- Applications per Job = DIVIDE([Total Applications], [Total Job Listings])
- Average Salary = AVERAGE('jobs'[simulatedSalary])
- Highest Salary = MAX('jobs'[simulatedSalary])

These DAX queries powered key card visuals and interactive metrics across all pages of the dashboard.