

# Data Wrangling Lab

Estimated time needed: **45** minutes

In this lab, you will perform data wrangling tasks to prepare raw data for analysis. Data wrangling involves cleaning, transforming, and organizing data into a structured format suitable for analysis. This lab focuses on tasks like identifying inconsistencies, encoding categorical variables, and feature transformation.

## Objectives

After completing this lab, you will be able to:

- Identify and remove inconsistent data entries.
- Encode categorical variables for analysis.
- Handle missing values using multiple imputation strategies.
- Apply feature scaling and transformation techniques.

## Install the required libraries

```
In [1]: !pip install pandas
        !pip install matplotlib
```

```
Requirement already satisfied: pandas in /opt/conda/lib/python3.12/site-packages (2.3.0)
Requirement already satisfied: numpy>=1.26.0 in /opt/conda/lib/python3.12/site-packages (from pandas) (2.3.0)
Requirement already satisfied: python-dateutil>=2.8.2 in /opt/conda/lib/python3.12/site-packages (from pandas) (2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in /opt/conda/lib/python3.12/site-packages (from pandas) (2024.2)
Requirement already satisfied: tzdata>=2022.7 in /opt/conda/lib/python3.12/site-packages (from pandas) (2025.2)
Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.12/site-packages (from python-dateutil>=2.8.2->pandas) (1.17.0)
Requirement already satisfied: matplotlib in /opt/conda/lib/python3.12/site-packages (3.10.3)
Requirement already satisfied: contourpy>=1.0.1 in /opt/conda/lib/python3.12/site-packages (from matplotlib) (1.3.2)
Requirement already satisfied: cycler>=0.10 in /opt/conda/lib/python3.12/site-packages (from matplotlib) (0.12.1)
Requirement already satisfied: fonttools>=4.22.0 in /opt/conda/lib/python3.12/site-packages (from matplotlib) (4.58.4)
Requirement already satisfied: kiwisolver>=1.3.1 in /opt/conda/lib/python3.12/site-packages (from matplotlib) (1.4.8)
Requirement already satisfied: numpy>=1.23 in /opt/conda/lib/python3.12/site-packages (from matplotlib) (2.3.0)
Requirement already satisfied: packaging>=20.0 in /opt/conda/lib/python3.12/site-packages (from matplotlib) (24.2)
Requirement already satisfied: pillow>=8 in /opt/conda/lib/python3.12/site-packages (from matplotlib) (11.2.1)
Requirement already satisfied: pyparsing>=2.3.1 in /opt/conda/lib/python3.12/site-packages (from matplotlib) (3.2.3)
Requirement already satisfied: python-dateutil>=2.7 in /opt/conda/lib/python3.12/site-packages (from matplotlib) (2.9.0.post0)
Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.12/site-packages (from python-dateutil>=2.7->matplotlib) (1.17.0)
```

# Tasks

Step 1: Import the necessary module.

## 1. Load the Dataset

1.1 Import necessary libraries and load the dataset.

Ensure the dataset is loaded correctly by displaying the first few rows.

```
In [2]: # Import necessary libraries
import pandas as pd

# Load the Stack Overflow survey data
dataset_url = "https://cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud/n01PQ9pSmiRX6520f
df = pd.read_csv(dataset_url)

# Display the first few rows
print(df.head())
```

	ResponseId	MainBranch	Age	\
0	1	I am a developer by profession	Under 18 years old	
1	2	I am a developer by profession	35-44 years old	
2	3	I am a developer by profession	45-54 years old	
3	4	I am learning to code	18-24 years old	
4	5	I am a developer by profession	18-24 years old	

	Employment	RemoteWork	Check	\
0	Employed, full-time	Remote	Apples	
1	Employed, full-time	Remote	Apples	
2	Employed, full-time	Remote	Apples	
3	Student, full-time	NaN	Apples	
4	Student, full-time	NaN	Apples	

	CodingActivities	\
0	Hobby	
1	Hobby;Contribute to open-source projects;Other...	
2	Hobby;Contribute to open-source projects;Other...	
3	NaN	
4	NaN	

	EdLevel	\
0	Primary/elementary school	
1	Bachelor's degree (B.A., B.S., B.Eng., etc.)	
2	Master's degree (M.A., M.S., M.Eng., MBA, etc.)	
3	Some college/university study without earning ...	
4	Secondary school (e.g. American high school, G...	

	LearnCode	\
0	Books / Physical media	
1	Books / Physical media;Colleague;On the job tr...	
2	Books / Physical media;Colleague;On the job tr...	
3	Other online resources (e.g., videos, blogs, f...	
4	Other online resources (e.g., videos, blogs, f...	

	LearnCodeOnline	...	JobSatPoints_6	\
0	NaN	...	NaN	
1	Technical documentation;Blogs;Books;Written Tu...	...	0.0	
2	Technical documentation;Blogs;Books;Written Tu...	...	NaN	
3	Stack Overflow;How-to videos;Interactive tutorial	...	NaN	
4	Technical documentation;Blogs;Written Tutorial...	...	NaN	

	JobSatPoints_7	JobSatPoints_8	JobSatPoints_9	JobSatPoints_10	\
0	NaN	NaN	NaN	NaN	
1	0.0	0.0	0.0	0.0	
2	NaN	NaN	NaN	NaN	
3	NaN	NaN	NaN	NaN	
4	NaN	NaN	NaN	NaN	

	JobSatPoints_11	SurveyLength	SurveyEase	ConvertedCompYearly	JobSat
0	NaN	NaN	NaN	NaN	NaN
1	0.0	NaN	NaN	NaN	NaN
2	NaN	Appropriate in length	Easy	NaN	NaN
3	NaN	Too long	Easy	NaN	NaN
4	NaN	Too short	Easy	NaN	NaN

[5 rows x 114 columns]

## 2. Explore the Dataset

2.1 Summarize the dataset by displaying the column data types, counts, and missing values.

```
In [10]: # Write your code here

print("---- Dataset Summary (df.dtypes) ----")
print(df.dtypes)

print("---- Dataset Summary (df.info()) ----")
df.info()

print("\n--- Missing Values Count (df.isnull().sum()) ---")
print(df.isnull().sum())
```

```

--- Dataset Summary (df.dtypes) ---
ResponseId      int64
MainBranch      object
Age             object
Employment      object
RemoteWork      object
...
JobSatPoints_11 float64
SurveyLength    object
SurveyEase      object
ConvertedCompYearly float64
JobSat          float64
Length: 114, dtype: object
--- Dataset Summary (df.info()) ---
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 65437 entries, 0 to 65436
Columns: 114 entries, ResponseId to JobSat
dtypes: float64(13), int64(1), object(100)
memory usage: 56.9+ MB

--- Missing Values Count (df.isnull().sum()) ---
ResponseId      0
MainBranch      0
Age             0
Employment      0
RemoteWork      10631
...
JobSatPoints_11 35992
SurveyLength    9255
SurveyEase      9199
ConvertedCompYearly 42002
JobSat          36311
Length: 114, dtype: int64

```

## 2.2 Generate basic statistics for numerical columns.

```

In [9]: # Write your code here

print("\n--- Summary Statistics (df.describe()) ---")
print(df.describe())

```

```

--- Summary Statistics (df.describe()) ---

```

	ResponseId	CompTotal	WorkExp	JobSatPoints_1	\
count	65437.000000	3.374000e+04	29658.000000	29324.000000	
mean	32719.000000	2.963841e+145	11.466957	18.581094	
std	18890.179119	5.444117e+147	9.168709	25.966221	
min	1.000000	0.000000e+00	0.000000	0.000000	
25%	16360.000000	6.000000e+04	4.000000	0.000000	
50%	32719.000000	1.100000e+05	9.000000	10.000000	
75%	49078.000000	2.500000e+05	16.000000	22.000000	
max	65437.000000	1.000000e+150	50.000000	100.000000	

	JobSatPoints_4	JobSatPoints_5	JobSatPoints_6	JobSatPoints_7	\
count	29393.000000	29411.000000	29450.000000	29448.00000	
mean	7.522140	10.060857	24.343232	22.96522	
std	18.422661	21.833836	27.089360	27.01774	
min	0.000000	0.000000	0.000000	0.00000	
25%	0.000000	0.000000	0.000000	0.00000	
50%	0.000000	0.000000	20.000000	15.00000	
75%	5.000000	10.000000	30.000000	30.00000	
max	100.000000	100.000000	100.000000	100.00000	

	JobSatPoints_8	JobSatPoints_9	JobSatPoints_10	JobSatPoints_11	\
count	29456.000000	29456.000000	29450.000000	29445.000000	
mean	20.278165	16.169432	10.955713	9.953948	
std	26.108110	24.845032	22.906263	21.775652	
min	0.000000	0.000000	0.000000	0.000000	
25%	0.000000	0.000000	0.000000	0.000000	
50%	10.000000	5.000000	0.000000	0.000000	
75%	25.000000	20.000000	10.000000	10.000000	
max	100.000000	100.000000	100.000000	100.000000	

	ConvertedCompYearly	JobSat
count	2.343500e+04	29126.000000
mean	8.615529e+04	6.935041
std	1.867570e+05	2.088259
min	1.000000e+00	0.000000
25%	3.271200e+04	6.000000
50%	6.500000e+04	7.000000
75%	1.079715e+05	8.000000
max	1.625660e+07	10.000000

### 3. Identifying and Removing Inconsistencies

3.1 Identify inconsistent or irrelevant entries in specific columns (e.g., Country).

```

In [11]: # Write your code here
import pandas as pd
import numpy as np

print("---- Identifying Inconsistent or Irrelevant Entries in 'Country' Column ----")

# 1. Display all unique values in the 'Country' column
print("\nUnique values in 'Country' column (before cleaning):")
print(df['Country'].unique())

# 2. Display value counts to see frequencies and spot inconsistencies
print("\nValue counts for 'Country' column (before cleaning):")
print(df['Country'].value_counts())

# 3. (Optional but recommended) Standardize text data for better identification
# Convert to lowercase and strip whitespace to catch common inconsistencies
df['Country_Cleaned'] = df['Country'].astype(str).str.lower().str.strip()

print("\nUnique values in 'Country' column (after basic cleaning - lowercase, strip whitespace):")
print(df['Country_Cleaned'].unique())
print("\nValue counts for 'Country_Cleaned' column (after basic cleaning):")
print(df['Country_Cleaned'].value_counts())

# 4. Identify irrelevant entries (e.g., 'N/A', placeholder strings, or clearly wrong entries)
# This step is highly dependent on your knowledge of the expected data.
irrelevant_entries = ['n/a', 'unknown', 'not applicable'] # Define what you consider irrelevant

print("\nIdentifying irrelevant entries (e.g., 'n/a', 'unknown'):")

```

```

for entry in irrelevant_entries:
    if entry in df['Country_Cleaned'].unique():
        print(f"Found irrelevant entry: '{entry}'")

# Example of how to address them (e.g., replace with NaN or remove)
# df['Country_Cleaned'] = df['Country_Cleaned'].replace(irrelevant_entries, np.nan)
# print("\n'Country_Cleaned' after replacing irrelevant entries with NaN:")
# print(df['Country_Cleaned'].unique())

# You would then decide how to handle these inconsistencies (e.g., correct typos,
# map similar values to a single standard value, or remove irrelevant entries).
# Example mapping:
# country_mapping = {
#     'usa': 'United States',
#     'u.s.a.': 'United States',
#     'uk': 'United Kingdom',
#     'canada ': 'Canada' # Already handled by strip(), but good for explicit mapping
# }
# df['Country_Cleaned'] = df['Country_Cleaned'].replace(country_mapping)
# print("\nUnique values after mapping common inconsistencies:")
# print(df['Country_Cleaned'].unique())

```

--- Identifying Inconsistent or Irrelevant Entries in 'Country' Column ---

Unique values in 'Country' column (before cleaning):

```
['United States of America'
 'United Kingdom of Great Britain and Northern Ireland' 'Canada' 'Norway'
 'Uzbekistan' 'Serbia' 'Poland' 'Philippines' 'Bulgaria' 'Switzerland'
 'India' 'Germany' 'Ireland' 'Italy' 'Ukraine' 'Australia' 'Brazil'
 'Japan' 'Austria' 'Iran, Islamic Republic of...' 'France' 'Saudi Arabia'
 'Romania' 'Turkey' 'Nepal' 'Algeria' 'Sweden' 'Netherlands' 'Croatia'
 'Pakistan' 'Czech Republic' 'Republic of North Macedonia' 'Finland'
 'Slovakia' 'Russian Federation' 'Greece' 'Israel' 'Belgium' 'Mexico'
 'United Republic of Tanzania' 'Hungary' 'Argentina' 'Portugal'
 'Sri Lanka' 'Latvia' 'China' 'Singapore' 'Lebanon' 'Spain' 'South Africa'
 'Lithuania' 'Viet Nam' 'Dominican Republic' 'Indonesia' 'Kosovo'
 'Morocco' 'Taiwan' 'Georgia' 'San Marino' 'Tunisia' 'Bangladesh'
 'Nigeria' 'Liechtenstein' 'Denmark' 'Ecuador' 'Malaysia' 'Albania'
 'Azerbaijan' 'Chile' 'Ghana' 'Peru' 'Bolivia' 'Egypt' 'Luxembourg'
 'Montenegro' 'Cyprus' 'Paraguay' 'Kazakhstan' 'Slovenia' 'Jordan'
 'Venezuela, Bolivarian Republic of...' 'Costa Rica' 'Jamaica' 'Thailand'
 'Nicaragua' 'Myanmar' 'Republic of Korea' 'Rwanda'
 'Bosnia and Herzegovina' 'Benin' 'El Salvador' 'Zimbabwe' 'Afghanistan'
 'Estonia' 'Malta' 'Uruguay' 'Belarus' 'Colombia' 'Republic of Moldova'
 'Isle of Man' 'Nomadic' 'New Zealand' 'Palestine' 'Armenia'
 'United Arab Emirates' 'Maldives' 'Ethiopia' 'Fiji' 'Guatemala' 'Uganda'
 'Turkmenistan' 'Mauritius' 'Kenya' 'Cuba' 'Gabon' 'Bahamas' 'South Korea'
 'Iceland' 'Honduras' 'Hong Kong (S.A.R.)'
 "Lao People's Democratic Republic" 'Mongolia' 'Cambodia' 'Madagascar'
 'Angola' 'Democratic Republic of the Congo' 'Syrian Arab Republic' 'Iraq'
 'Namibia' 'Senegal' 'Kyrgyzstan' 'Zambia' 'Swaziland' "Côte d'Ivoire"
 'Kuwait' 'Tajikistan' 'Burundi' 'Trinidad and Tobago' 'Mauritania'
 'Sierra Leone' 'Panama' 'Somalia' 'North Korea' 'Dominica' 'Guyana'
 'Togo' 'Oman' 'Barbados' 'Andorra'
 "Democratic People's Republic of Korea" 'Qatar' 'Sudan' 'Cameroon'
 'Papua New Guinea' 'Bahrain' 'Yemen' 'Malawi' 'Burkina Faso'
 'Congo, Republic of the...' 'Botswana' 'Guinea-Bissau' 'Mozambique'
 'Central African Republic' 'Equatorial Guinea' 'Suriname' 'Belize'
 'Libyan Arab Jamahiriya' 'Cape Verde' 'Brunei Darussalam' 'Bhutan'
 'Guinea' 'Niger' 'Antigua and Barbuda' 'Mali' 'Samoa' 'Lesotho'
 'Saint Kitts and Nevis' 'Monaco' 'Micronesia, Federated States of...'
 'Haiti' nan 'Nauru' 'Liberia' 'Chad' 'Djibouti' 'Solomon Islands']
```

Value counts for 'Country' column (before cleaning):

```
Country
United States of America      11095
Germany                      4947
India                        4231
United Kingdom of Great Britain and Northern Ireland  3224
Ukraine                      2672
...
Micronesia, Federated States of...  1
Nauru                          1
Chad                           1
Djibouti                       1
Solomon Islands                1
Name: count, Length: 185, dtype: int64
```

Unique values in 'Country' column (after basic cleaning - lowercase, strip whitespace):

```
['united states of america'
 'united kingdom of great britain and northern ireland' 'canada' 'norway'
 'uzbekistan' 'serbia' 'poland' 'philippines' 'bulgaria' 'switzerland'
 'india' 'germany' 'ireland' 'italy' 'ukraine' 'australia' 'brazil'
 'japan' 'austria' 'iran, islamic republic of...' 'france' 'saudi arabia'
 'romania' 'turkey' 'nepal' 'algeria' 'sweden' 'netherlands' 'croatia'
 'pakistan' 'czech republic' 'republic of north macedonia' 'finland'
 'slovakia' 'russian federation' 'greece' 'israel' 'belgium' 'mexico'
 'united republic of tanzania' 'hungary' 'argentina' 'portugal'
 'sri lanka' 'latvia' 'china' 'singapore' 'lebanon' 'spain' 'south africa'
 'lithuania' 'viet nam' 'dominican republic' 'indonesia' 'kosovo'
 'morocco' 'taiwan' 'georgia' 'san marino' 'tunisia' 'bangladesh'
 'nigeria' 'liechtenstein' 'denmark' 'ecuador' 'malaysia' 'albania'
 'azerbaijan' 'chile' 'ghana' 'peru' 'bolivia' 'egypt' 'luxembourg'
 'montenegro' 'cyprus' 'paraguay' 'kazakhstan' 'slovenia' 'jordan'
 'venezuela, bolivarian republic of...' 'costa rica' 'jamaica' 'thailand'
 'nicaragua' 'myanmar' 'republic of korea' 'rwanda'
 'bosnia and herzegovina' 'benin' 'el salvador' 'zimbabwe' 'afghanistan']
```

'estonia' 'malta' 'uruguay' 'belarus' 'colombia' 'republic of moldova'  
 'isle of man' 'nomadic' 'new zealand' 'palestine' 'armenia'  
 'united arab emirates' 'maldives' 'ethiopia' 'fiji' 'guatemala' 'uganda'  
 'turkmenistan' 'mauritius' 'kenya' 'cuba' 'gabon' 'bahamas' 'south korea'  
 'iceland' 'honduras' 'hong kong (s.a.r.)'  
 'lao people's democratic republic' 'mongolia' 'cambodia' 'madagascar'  
 'angola' 'democratic republic of the congo' 'syrian arab republic' 'iraq'  
 'namibia' 'senegal' 'kyrgyzstan' 'zambia' 'swaziland' 'côte d'ivoire'  
 'kuwait' 'tajikistan' 'burundi' 'trinidad and tobago' 'mauritania'  
 'sierra leone' 'panama' 'somalia' 'north korea' 'dominica' 'guyana'  
 'togo' 'oman' 'barbados' 'andorra'  
 'democratic people's republic of korea' 'qatar' 'sudan' 'cameroon'  
 'papua new guinea' 'bahrain' 'yemen' 'malawi' 'burkina faso'  
 'congo, republic of the...' 'botswana' 'guinea-bissau' 'mozambique'  
 'central african republic' 'equatorial guinea' 'suriname' 'belize'  
 'libyan arab jamahiriya' 'cape verde' 'brunei darussalam' 'bhutan'  
 'guinea' 'niger' 'antigua and barbuda' 'mali' 'samoa' 'lesotho'  
 'saint kitts and nevis' 'monaco' 'micronesia, federated states of...'  
 'haiti' 'nan' 'nauru' 'liberia' 'chad' 'djibouti' 'solomon islands']

Value counts for 'Country\_Cleaned' column (after basic cleaning):

Country_Cleaned	
united states of america	11095
nan	6507
germany	4947
india	4231
united kingdom of great britain and northern ireland	3224
	...
micronesia, federated states of...	1
nauru	1
chad	1
djibouti	1
solomon islands	1

Name: count, Length: 186, dtype: int64

Identifying irrelevant entries (e.g., 'n/a', 'unknown'):

3.2 Standardize entries in columns like Country or EdLevel by mapping inconsistent values to a consistent format.

```
In [12]: ## Write your code here
# --- New Section: 3.2 Standardize Entries in Categorical Columns ---
print("\n--- 3.2 Standardize entries in columns like Country or EdLevel by mapping inconsistent val

# --- Standardizing 'Country' Column ---
print("\nStandardizing 'Country' column:")
print("Original unique values in 'Country':", df['Country'].unique())

# Step 1: Convert to string, lowercase, and strip whitespace for initial consistency
df['Country'] = df['Country'].astype(str).str.lower().str.strip()

# Step 2: Define mapping for common inconsistencies
country_mapping = {
    'usa': 'United States',
    'u.s.a.': 'United States',
    'uk': 'United Kingdom',
    'n/a': np.nan # Treat 'N/A' as a missing value
}
df['Country'] = df['Country'].replace(country_mapping)

print("Unique values in 'Country' after standardization:")
print(df['Country'].unique())
print("Value counts for 'Country' after standardization:")
print(df['Country'].value_counts(dropna=False)) # Show counts including NaN if any

# --- Standardizing 'EdLevel' Column ---
print("\nStandardizing 'EdLevel' column:")
print("Original unique values in 'EdLevel':", df['EdLevel'].unique())

# Step 1: Convert to string, lowercase, and strip whitespace
df['EdLevel'] = df['EdLevel'].astype(str).str.lower().str.strip()

# Step 2: Define mapping for common inconsistencies
edlevel_mapping = {
    'some college': 'Some college/university study without earning a degree',
```



```

    'masters degree': 'Masters degree',
    'phd': 'PhD',
    'high school': 'Less than a Bachelors degree', # Assuming this maps to 'Less than a Bachelor's'
    # Add more mappings as needed based on your dataset's specific inconsistencies
}
df['EdLevel'] = df['EdLevel'].replace(edlevel_mapping)

# Capitalize first letter of each word if desired for presentation
df['EdLevel'] = df['EdLevel'].str.title()

print("Unique values in 'EdLevel' after standardization:")
print(df['EdLevel'].unique())
print("Value counts for 'EdLevel' after standardization:")
print(df['EdLevel'].value_counts(dropna=False)) # Show counts including NaN if any

print("\nStandardization of categorical columns 'Country' and 'EdLevel' complete.")
print("This process helps in ensuring data consistency for accurate analysis and visualization.")

```

--- 3.2 Standardize entries in columns like Country or EdLevel by mapping inconsistent values to a consistent format ---

Standardizing 'Country' column:

Original unique values in 'Country': ['United States of America'  
'United Kingdom of Great Britain and Northern Ireland' 'Canada' 'Norway'  
'Uzbekistan' 'Serbia' 'Poland' 'Philippines' 'Bulgaria' 'Switzerland'  
'India' 'Germany' 'Ireland' 'Italy' 'Ukraine' 'Australia' 'Brazil'  
'Japan' 'Austria' 'Iran, Islamic Republic of...' 'France' 'Saudi Arabia'  
'Romania' 'Turkey' 'Nepal' 'Algeria' 'Sweden' 'Netherlands' 'Croatia'  
'Pakistan' 'Czech Republic' 'Republic of North Macedonia' 'Finland'  
'Slovakia' 'Russian Federation' 'Greece' 'Israel' 'Belgium' 'Mexico'  
'United Republic of Tanzania' 'Hungary' 'Argentina' 'Portugal'  
'Sri Lanka' 'Latvia' 'China' 'Singapore' 'Lebanon' 'Spain' 'South Africa'  
'Lithuania' 'Viet Nam' 'Dominican Republic' 'Indonesia' 'Kosovo'  
'Morocco' 'Taiwan' 'Georgia' 'San Marino' 'Tunisia' 'Bangladesh'  
'Nigeria' 'Liechtenstein' 'Denmark' 'Ecuador' 'Malaysia' 'Albania'  
'Azerbaijan' 'Chile' 'Ghana' 'Peru' 'Bolivia' 'Egypt' 'Luxembourg'  
'Montenegro' 'Cyprus' 'Paraguay' 'Kazakhstan' 'Slovenia' 'Jordan'  
'Venezuela, Bolivarian Republic of...' 'Costa Rica' 'Jamaica' 'Thailand'  
'Nicaragua' 'Myanmar' 'Republic of Korea' 'Rwanda'  
'Bosnia and Herzegovina' 'Benin' 'El Salvador' 'Zimbabwe' 'Afghanistan'  
'Estonia' 'Malta' 'Uruguay' 'Belarus' 'Colombia' 'Republic of Moldova'  
'Isle of Man' 'Nomadic' 'New Zealand' 'Palestine' 'Armenia'  
'United Arab Emirates' 'Maldives' 'Ethiopia' 'Fiji' 'Guatemala' 'Uganda'  
'Turkmenistan' 'Mauritius' 'Kenya' 'Cuba' 'Gabon' 'Bahamas' 'South Korea'  
'Iceland' 'Honduras' 'Hong Kong (S.A.R.)'  
'Lao People's Democratic Republic' 'Mongolia' 'Cambodia' 'Madagascar'  
'Angola' 'Democratic Republic of the Congo' 'Syrian Arab Republic' 'Iraq'  
'Namibia' 'Senegal' 'Kyrgyzstan' 'Zambia' 'Swaziland' 'Côte d'Ivoire'  
'Kuwait' 'Tajikistan' 'Burundi' 'Trinidad and Tobago' 'Mauritania'  
'Sierra Leone' 'Panama' 'Somalia' 'North Korea' 'Dominica' 'Guyana'  
'Togo' 'Oman' 'Barbados' 'Andorra'  
'Democratic People's Republic of Korea' 'Qatar' 'Sudan' 'Cameroon'  
'Papua New Guinea' 'Bahrain' 'Yemen' 'Malawi' 'Burkina Faso'  
'Congo, Republic of the...' 'Botswana' 'Guinea-Bissau' 'Mozambique'  
'Central African Republic' 'Equatorial Guinea' 'Suriname' 'Belize'  
'Libyan Arab Jamahiriya' 'Cape Verde' 'Brunei Darussalam' 'Bhutan'  
'Guinea' 'Niger' 'Antigua and Barbuda' 'Mali' 'Samoa' 'Lesotho'  
'Saint Kitts and Nevis' 'Monaco' 'Micronesia, Federated States of...'  
'Haiti' nan 'Nauru' 'Liberia' 'Chad' 'Djibouti' 'Solomon Islands']

Unique values in 'Country' after standardization:

['united states of america'  
'united kingdom of great britain and northern ireland' 'canada' 'norway'  
'uzbekistan' 'serbia' 'poland' 'philippines' 'bulgaria' 'switzerland'  
'india' 'germany' 'ireland' 'italy' 'ukraine' 'australia' 'brazil'  
'japan' 'austria' 'iran, islamic republic of...' 'france' 'saudi arabia'  
'romania' 'turkey' 'nepal' 'algeria' 'sweden' 'netherlands' 'croatia'  
'pakistan' 'czech republic' 'republic of north macedonia' 'finland'  
'slovakia' 'russian federation' 'greece' 'israel' 'belgium' 'mexico'  
'united republic of tanzania' 'hungary' 'argentina' 'portugal'  
'sri lanka' 'latvia' 'china' 'singapore' 'lebanon' 'spain' 'south africa'  
'lithuania' 'viet nam' 'dominican republic' 'indonesia' 'kosovo'  
'morocco' 'taiwan' 'georgia' 'san marino' 'tunisia' 'bangladesh'  
'nigeria' 'liechtenstein' 'denmark' 'ecuador' 'malaysia' 'albania'  
'azerbaijan' 'chile' 'ghana' 'peru' 'bolivia' 'egypt' 'luxembourg'  
'montenegro' 'cyprus' 'paraguay' 'kazakhstan' 'slovenia' 'jordan'  
'venezuela, bolivarian republic of...' 'costa rica' 'jamaica' 'thailand'  
'nicaragua' 'myanmar' 'republic of korea' 'rwanda'  
'bosnia and herzegovina' 'benin' 'el salvador' 'zimbabwe' 'afghanistan'  
'estonia' 'malta' 'uruguay' 'belarus' 'colombia' 'republic of moldova'  
'isle of man' 'nomadic' 'new zealand' 'palestine' 'armenia'  
'united arab emirates' 'maldives' 'ethiopia' 'fiji' 'guatemala' 'uganda'  
'turkmenistan' 'mauritius' 'kenya' 'cuba' 'gabon' 'bahamas' 'south korea'  
'iceland' 'honduras' 'hong kong (s.a.r.)'  
'lao people's democratic republic' 'mongolia' 'cambodia' 'madagascar'  
'angola' 'democratic republic of the congo' 'syrian arab republic' 'iraq'  
'namibia' 'senegal' 'kyrgyzstan' 'zambia' 'swaziland' 'côte d'ivoire'  
'kuwait' 'tajikistan' 'burundi' 'trinidad and tobago' 'mauritania'  
'sierra leone' 'panama' 'somalia' 'north korea' 'dominica' 'guyana'  
'togo' 'oman' 'barbados' 'andorra'  
'democratic people's republic of korea' 'qatar' 'sudan' 'cameroon'  
'papua new guinea' 'bahrain' 'yemen' 'malawi' 'burkina faso'  
'congo, republic of the...' 'botswana' 'guinea-bissau' 'mozambique'  
'central african republic' 'equatorial guinea' 'suriname' 'belize']

```
'libyan arab jamahiriya' 'cape verde' 'brunei darussalam' 'bhutan'
'guinea' 'niger' 'antigua and barbuda' 'mali' 'samoa' 'lesotho'
'saint kitts and nevis' 'monaco' 'micronesia, federated states of...'
'haiti' 'nan' 'nauru' 'liberia' 'chad' 'djibouti' 'solomon islands']
```

Value counts for 'Country' after standardization:

```
Country
united states of america    11095
nan                          6507
germany                     4947
india                       4231
united kingdom of great britain and northern ireland    3224
...
micronesia, federated states of...    1
nauru                                1
chad                                1
djibouti                            1
solomon islands                    1
```

Name: count, Length: 186, dtype: int64

Standardizing 'EdLevel' column:

Original unique values in 'EdLevel': ['Primary/elementary school'

'Bachelor's degree (B.A., B.S., B.Eng., etc.)'

'Master's degree (M.A., M.S., M.Eng., MBA, etc.)'

'Some college/university study without earning a degree'

'Secondary school (e.g. American high school, German Realschule or Gymnasium, etc.)'

'Professional degree (JD, MD, Ph.D, Ed.D, etc.)'

'Associate degree (A.A., A.S., etc.)' 'Something else' nan]

Unique values in 'EdLevel' after standardization:

['Primary/Elementary School'

'Bachelor'S Degree (B.A., B.S., B.Eng., Etc.)'

'Master'S Degree (M.A., M.S., M.Eng., Mba, Etc.)'

'Some College/University Study Without Earning A Degree'

'Secondary School (E.G. American High School, German Realschule Or Gymnasium, Etc.)'

'Professional Degree (Jd, Md, Ph.D, Ed.D, Etc.)'

'Associate Degree (A.A., A.S., Etc.)' 'Something Else' 'Nan']

Value counts for 'EdLevel' after standardization:

```
EdLevel
Bachelor'S Degree (B.A., B.S., B.Eng., Etc.)    24942
Master'S Degree (M.A., M.S., M.Eng., Mba, Etc.)    15557
Some College/University Study Without Earning A Degree    7651
Secondary School (E.G. American High School, German Realschule Or Gymnasium, Etc.)    5793
Nan    4653
Professional Degree (Jd, Md, Ph.D, Ed.D, Etc.)    2970
Associate Degree (A.A., A.S., Etc.)    1793
Primary/Elementary School    1146
Something Else    932
```

Name: count, dtype: int64

Standardization of categorical columns 'Country' and 'EdLevel' complete.

This process helps in ensuring data consistency for accurate analysis and visualization.

## 4. Encoding Categorical Variables

### 4.1 Encode the Employment column using one-hot encoding.

```
In [13]: ## Write your code here
# --- New Section: 4.1 Encode the Employment column using one-hot encoding ---
print("\n--- 4.1 Encode the Employment column using one-hot encoding ---")

# Display original Employment column info and head
print("\nOriginal 'Employment' column head:")
print(df['Employment'].head())
print("\nOriginal 'Employment' unique values:")
print(df['Employment'].unique())

# Perform one-hot encoding on the 'Employment' column
# drop_first=True is often used to avoid multicollinearity if you're using this for modeling.
# It drops one of the generated columns, as the information is implicitly contained in the others.
employment_encoded = pd.get_dummies(df['Employment'], prefix='Employment', drop_first=False)

# Concatenate the new one-hot encoded columns with the original DataFrame
df = pd.concat([df, employment_encoded], axis=1)
```

```
# Drop the original 'Employment' column if you no longer need it
# df = df.drop('Employment', axis=1)

print("\nDataFrame head after one-hot encoding 'Employment' column (showing new columns):")
print(df.head())

print(f"\nNew columns created from 'Employment' encoding: {employment_encoded.columns.tolist()}")
print("\nShape of DataFrame after one-hot encoding:", df.shape)
print("One-hot encoding of 'Employment' column complete. This converts categorical data into a nume
```

--- 4.1 Encode the Employment column using one-hot encoding ---

Original 'Employment' column head:

```
0    Employed, full-time
1    Employed, full-time
2    Employed, full-time
3    Student, full-time
4    Student, full-time
```

Name: Employment, dtype: object

Original 'Employment' unique values:

```
['Employed, full-time' 'Student, full-time'
 'Student, full-time;Not employed, but looking for work'
 'Independent contractor, freelancer, or self-employed'
 'Not employed, and not looking for work'
 'Employed, full-time;Student, part-time'
 'Employed, full-time;Independent contractor, freelancer, or self-employed'
 'Employed, full-time;Student, full-time' 'Employed, part-time'
 'Student, full-time;Employed, part-time'
 'Student, part-time;Employed, part-time' 'I prefer not to say'
 'Not employed, but looking for work' 'Student, part-time'
 'Employed, full-time;Student, full-time;Independent contractor, freelancer, or self-employed;Employ
ed, part-time'
 'Employed, full-time;Independent contractor, freelancer, or self-employed;Student, part-time'
 'Independent contractor, freelancer, or self-employed;Employed, part-time'
 'Independent contractor, freelancer, or self-employed;Student, part-time;Employed, part-time'
 'Student, full-time;Not employed, but looking for work;Independent contractor, freelancer, or self-
employed'
 'Student, full-time;Independent contractor, freelancer, or self-employed'
 'Employed, full-time;Employed, part-time'
 'Not employed, but looking for work;Independent contractor, freelancer, or self-employed'
 'Student, full-time;Not employed, and not looking for work' 'Retired'
 'Independent contractor, freelancer, or self-employed;Student, part-time'
 'Employed, full-time;Independent contractor, freelancer, or self-employed;Employed, part-time'
 'Not employed, but looking for work;Independent contractor, freelancer, or self-employed;Student, p
art-time'
 'Not employed, but looking for work;Student, part-time'
 'Not employed, but looking for work;Not employed, and not looking for work'
 'Independent contractor, freelancer, or self-employed;Retired'
 'Not employed, but looking for work;Student, part-time;Employed, part-time'
 'Student, full-time;Not employed, but looking for work;Not employed, and not looking for work'
 'Employed, full-time;Not employed, but looking for work'
 'Student, full-time;Not employed, and not looking for work;Student, part-time'
 'Employed, full-time;Retired'
 'Employed, full-time;Independent contractor, freelancer, or self-employed;Student, part-time;Employ
ed, part-time'
 'Not employed, but looking for work;Independent contractor, freelancer, or self-employed;Not employ
ed, and not looking for work'
 'Not employed, but looking for work;Independent contractor, freelancer, or self-employed;Employed,
part-time'
 'Not employed, but looking for work;Employed, part-time'
 'Employed, full-time;Student, full-time;Employed, part-time'
 'Independent contractor, freelancer, or self-employed;Not employed, and not looking for work'
 'Not employed, and not looking for work;Student, part-time'
 'Student, full-time;Independent contractor, freelancer, or self-employed;Employed, part-time'
 'Student, full-time;Student, part-time'
 'Student, full-time;Not employed, but looking for work;Student, part-time'
 'Independent contractor, freelancer, or self-employed;Not employed, and not looking for work;Retire
d'
 'Employed, full-time;Independent contractor, freelancer, or self-employed;Not employed, and not loo
king for work'
 'Employed, full-time;Student, full-time;Independent contractor, freelancer, or self-employed'
 'Employed, full-time;Student, full-time;Student, part-time'
 'Not employed, but looking for work;Retired'
 'Employed, full-time;Student, full-time;Not employed, but looking for work'
 'Not employed, and not looking for work;Retired'
 'Not employed, but looking for work;Independent contractor, freelancer, or self-employed;Not employ
ed, and not looking for work;Retired'
 'Employed, full-time;Not employed, but looking for work;Employed, part-time'
 'Student, full-time;Not employed, but looking for work;Independent contractor, freelancer, or self-
employed;Student, part-time;Employed, part-time;Retired'
 'Employed, full-time;Independent contractor, freelancer, or self-employed;Not employed, and not loo
king for work;Employed, part-time'
 'Student, full-time;Independent contractor, freelancer, or self-employed;Not employed, and not look
ing for work'
```

[illegible]

'Student, full-time;Not employed, but looking for work;Independent contractor, freelancer, or self-employed;Not employed, and not looking for work;Student, part-time'  
 'Employed, full-time;Student, full-time;Not employed, but looking for work;Independent contractor, freelancer, or self-employed;Student, part-time;Employed, part-time;Retired'  
 'Not employed, but looking for work;Independent contractor, freelancer, or self-employed;Not employed, and not looking for work;Employed, part-time'  
 'Student, full-time;Retired'  
 'Employed, full-time;Not employed, but looking for work;Student, part-time'  
 'Not employed, and not looking for work;Student, part-time;Employed, part-time'  
 'Not employed, but looking for work;Independent contractor, freelancer, or self-employed;Student, part-time;Retired']

DataFrame head after one-hot encoding 'Employment' column (showing new columns):

	ResponseId	MainBranch	Age \
0	1	I am a developer by profession	Under 18 years old
1	2	I am a developer by profession	35-44 years old
2	3	I am a developer by profession	45-54 years old
3	4	I am learning to code	18-24 years old
4	5	I am a developer by profession	18-24 years old

	Employment	RemoteWork	Check \
0	Employed, full-time	Remote	Apples
1	Employed, full-time	Remote	Apples
2	Employed, full-time	Remote	Apples
3	Student, full-time	NaN	Apples
4	Student, full-time	NaN	Apples

	CodingActivities \
0	Hobby
1	Hobby;Contribute to open-source projects;Other...
2	Hobby;Contribute to open-source projects;Other...
3	NaN
4	NaN

	EdLevel \
0	Primary/Elementary School
1	Bachelor'S Degree (B.A., B.S., B.Eng., Etc.)
2	Master'S Degree (M.A., M.S., M.Eng., Mba, Etc.)
3	Some College/University Study Without Earning ...
4	Secondary School (E.G. American High School, G...

	LearnCode \
0	Books / Physical media
1	Books / Physical media;Colleague;On the job tr...
2	Books / Physical media;Colleague;On the job tr...
3	Other online resources (e.g., videos, blogs, f...
4	Other online resources (e.g., videos, blogs, f...

	LearnCodeOnline ... \
0	NaN ...
1	Technical documentation;Blogs;Books;Written Tu... ...
2	Technical documentation;Blogs;Books;Written Tu... ...
3	Stack Overflow;How-to videos;Interactive tutorial ...
4	Technical documentation;Blogs;Written Tutorial... ...

	Employment_Student, full-time;Not employed, but looking for work;Not employed, and not looking for work;Student, part-time \
0	False
1	False
2	False
3	False
4	False

	Employment_Student, full-time;Not employed, but looking for work;Retired \
0	False
1	False
2	False
3	False
4	False

	Employment_Student, full-time;Not employed, but looking for work;Student, part-time \
0	False
1	False
2	False
3	False

	False
Employment_Student, full-time;Retired \	
0	False
1	False
2	False
3	False
4	False

Employment_Student, full-time;Student, part-time \	
0	False
1	False
2	False
3	False
4	False

Employment_Student, full-time;Student, part-time;Employed, part-time \	
0	False
1	False
2	False
3	False
4	False

Employment_Student, full-time;Student, part-time;Retired \	
0	False
1	False
2	False
3	False
4	False

Employment_Student, part-time \	
0	False
1	False
2	False
3	False
4	False

Employment_Student, part-time;Employed, part-time \	
0	False
1	False
2	False
3	False
4	False

Employment_Student, part-time;Retired	
0	False
1	False
2	False
3	False
4	False

[5 rows x 225 columns]

New columns created from 'Employment' encoding: ['Employment\_Employed, full-time', 'Employment\_Employed, full-time;Employed, part-time', 'Employment\_Employed, full-time;Independent contractor, freelancer, or self-employed', 'Employment\_Employed, full-time;Independent contractor, freelancer, or self-employed;Employed, part-time', 'Employment\_Employed, full-time;Independent contractor, freelancer, or self-employed;Employed, part-time;Retired', 'Employment\_Employed, full-time;Independent contractor, freelancer, or self-employed;Not employed, and not looking for work', 'Employment\_Employed, full-time;Independent contractor, freelancer, or self-employed;Not employed, and not looking for work;Employed, part-time', 'Employment\_Employed, full-time;Independent contractor, freelancer, or self-employed;Not employed, and not looking for work;Student, part-time', 'Employment\_Employed, full-time;Independent contractor, freelancer, or self-employed;Retired', 'Employment\_Employed, full-time;Independent contractor, freelancer, or self-employed;Student, part-time', 'Employment\_Employed, full-time;Independent contractor, freelancer, or self-employed;Student, part-time;Employed, part-time', 'Employment\_Employed, full-time;Independent contractor, freelancer, or self-employed;Student, part-time;Retired', 'Employment\_Employed, full-time;Not employed, and not looking for work', 'Employment\_Employed, full-time;Not employed, but looking for work', 'Employment\_Employed, full-time;Not employed, but looking for work;Employed, part-time', 'Employment\_Employed, full-time;Not employed, but looking for work;Independent contractor, freelancer, or self-employed', 'Employment\_Employed, full-time;Not employed, but looking for work;Independent contractor, freelancer, or self-employed;Employed, part-time', 'Employment\_Employed, full-time;Not employed, but looking for work;Independent contractor, freelancer, or self-employed;Not employed, and not looking for work;Employed, part-time', 'Employment\_Employed, full-time;Not employed, but looking for work;Not employed, and not looking for work;Employed, part-time', 'Employment\_Employed, full-time;Not employed, but looking for work;Student, part-time', 'Em



[illegible]

k;Independent contractor, freelancer, or self-employed;Employed, part-time', 'Employment\_Student, full-time;Not employed, but looking for work;Independent contractor, freelancer, or self-employed;Not employed, and not looking for work', 'Employment\_Student, full-time;Not employed, but looking for work;Independent contractor, freelancer, or self-employed;Not employed, and not looking for work;Student, part-time', 'Employment\_Student, full-time;Not employed, but looking for work;Independent contractor, freelancer, or self-employed;Student, part-time', 'Employment\_Student, full-time;Not employed, but looking for work;Independent contractor, freelancer, or self-employed;Student, part-time;Employed, part-time;Retired', 'Employment\_Student, full-time;Not employed, but looking for work;Not employed, and not looking for work', 'Employment\_Student, full-time;Not employed, but looking for work;Not employed, and not looking for work;Student, part-time', 'Employment\_Student, full-time;Not employed, but looking for work;Retired', 'Employment\_Student, full-time;Not employed, but looking for work;Student, part-time', 'Employment\_Student, full-time;Retired', 'Employment\_Student, full-time;Student, part-time', 'Employment\_Student, full-time;Student, part-time;Employed, part-time', 'Employment\_Student, full-time;Student, part-time;Retired', 'Employment\_Student, part-time', 'Employment\_Student, part-time;Employed, part-time', 'Employment\_Student, part-time;Retired']

Shape of DataFrame after one-hot encoding: (65437, 225)

One-hot encoding of 'Employment' column complete. This converts categorical data into a numerical format suitable for machine learning models.

## 5. Handling Missing Values

### 5.1 Identify columns with the highest number of missing values.

```
In [14]: ## Write your code here
# 5.1 Identify columns with the highest number of missing values.
print("\n5.1 Identify columns with the highest number of missing values.")
missing_values_count = df.isnull().sum()
# Filter to show only columns with missing values and sort them
columns_with_missing = missing_values_count[missing_values_count > 0].sort_values(ascending=False)

if not columns_with_missing.empty:
    print("\nColumns with missing values (highest first):")
    print(columns_with_missing)
    # Get the column with the absolute highest number of missing values
    highest_missing_column = columns_with_missing.index[0]
    print(f"\nColumn with the highest number of missing values: '{highest_missing_column}' ({columns_with_missing[highest_missing_column]:.2f})")
else:
    print("\nNo missing values found in the dataset.")
```

5.1 Identify columns with the highest number of missing values.

Columns with missing values (highest first):

AINextMuch less integrated	64289
AINextLess integrated	63082
AINextNo change	52939
AINextMuch more integrated	51999
EmbeddedAdmired	48704

LanguageHaveWorkedWith	5692
YearsCode	5568
NEWSOSites	5151
LearnCode	4949
AISelect	4530

Length: 107, dtype: int64

Column with the highest number of missing values: 'AINextMuch less integrated' (64289 missing)

### 5.2 Impute missing values in numerical columns (e.g., 'ConvertedCompYearly') with the mean or median.

```
In [15]: ## Write your code here
# 5.2 Impute missing values in numerical columns (e.g., 'ConvertedCompYearly') with the mean or median
print("\n5.2 Impute missing values in numerical columns with mean or median.")
# For 'ConvertedCompYearly', median is generally preferred for skewed data like income
# to be less affected by extreme outliers.
if 'ConvertedCompYearly' in df.columns and df['ConvertedCompYearly'].isnull().any():
    median_comp = df['ConvertedCompYearly'].median()
    df['ConvertedCompYearly'].fillna(median_comp, inplace=True)
    print(f"Missing values in 'ConvertedCompYearly' imputed with median: {median_comp:.2f}")
else:
    print("'ConvertedCompYearly' column not found or has no missing values. No imputation performed.")
```

5.2 Impute missing values in numerical columns with mean or median.

Missing values in 'ConvertedCompYearly' imputed with median: 65000.00

/tmp/ipykernel\_2452/3683777211.py:8: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.  
The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values always behaves as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform the operation inplace on the original object.

```
df['ConvertedCompYearly'].fillna(median_comp, inplace=True)
```

5.3 Impute missing values in categorical columns (e.g., 'RemoteWork') with the most frequent value.

```
In [16]: ## Write your code here
# 5.3 Impute missing values in categorical columns (e.g., RemoteWork) with the most frequent value.
print("\n5.3 Impute missing values in categorical columns with the most frequent value.")
if 'RemoteWork' in df.columns and df['RemoteWork'].isnull().any():
    most_frequent_remotework = df['RemoteWork'].mode()[0]
    df['RemoteWork'].fillna(most_frequent_remotework, inplace=True)
    print(f'Missing values in 'RemoteWork' imputed with most frequent value: '{most_frequent_remote
else:
    print("'RemoteWork' column not found or has no missing values. No imputation performed.")

# Verify missing values after handling
print("\nMissing values after Section 5 imputation:")
print(df.isnull().sum()[df.isnull().sum() > 0]) # Show only columns that still have NaNs
```

5.3 Impute missing values in categorical columns with the most frequent value.

Missing values in 'RemoteWork' imputed with most frequent value: 'Hybrid (some remote, some in-person)'

Missing values after Section 5 imputation:

/tmp/ipykernel\_2452/2099079762.py:6: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.  
The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values always behaves as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform the operation inplace on the original object.

```
df['RemoteWork'].fillna(most_frequent_remotework, inplace=True)
```

```
CodingActivities    10971
LearnCode           4949
LearnCodeOnline     16200
TechDoc             24540
YearsCode           5568
...
JobSatPoints_10     35987
JobSatPoints_11     35992
SurveyLength        9255
SurveyEase          9199
JobSat              36311
Length: 105, dtype: int64
```

## 6. Feature Scaling and Transformation

6.1 Apply Min-Max Scaling to normalize the 'ConvertedCompYearly' column.

```
In [24]: ## Write your code here

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.preprocessing import MinMaxScaler
from scipy.stats import skew

#6.1 Apply Min-Max Scaling to normalize the ConvertedCompYearly column.
print("\n6.1 Apply Min-Max Scaling to normalize the 'ConvertedCompYearly' column.")
if 'ConvertedCompYearly' in df.columns and pd.api.types.is_numeric_dtype(df['ConvertedCompYearly'])
```

```

scaler_minmax = MinMaxScaler()
# Assign to a NEW column to keep original, and match print statement's intention
df['ConvertedCompYearly_MinMax'] = scaler_minmax.fit_transform(df[['ConvertedCompYearly']])
print("Min-Max Scaling applied. New column 'ConvertedCompYearly_MinMax' created.")
print("Descriptive stats for 'ConvertedCompYearly_MinMax':")
# Corrected column name in .describe()
print(df['ConvertedCompYearly_MinMax'].describe())
else:
    print("'ConvertedCompYearly' not suitable for Min-Max Scaling (not found or not numeric).")

```

6.1 Apply Min-Max Scaling to normalize the 'ConvertedCompYearly' column.  
Min-Max Scaling applied. New column 'ConvertedCompYearly\_MinMax' created.  
Descriptive stats for 'ConvertedCompYearly\_MinMax':

count	65437.000000
mean	0.004464
std	0.006903
min	0.000000
25%	0.003998
50%	0.003998
75%	0.003998
max	1.000000

Name: ConvertedCompYearly\_MinMax, dtype: float64

6.2 Log-transform the ConvertedCompYearly column to reduce skewness.

```

In [19]: ## Write your code here
print("\n6.2 Log-transform the ConvertedCompYearly column to reduce skewness.")
if 'ConvertedCompYearly' in df.columns and pd.api.types.is_numeric_dtype(df['ConvertedCompYearly']):
    # Check skewness before transformation (optional)
    original_skew = skew(df['ConvertedCompYearly'].dropna())
    print(f"Original skewness of 'ConvertedCompYearly': {original_skew:.2f}")

    # Apply log transformation. Add 1 to handle zero or negative values if they exist, to avoid log
    df['ConvertedCompYearly_Log'] = np.log1p(df['ConvertedCompYearly'])
    print("Log transformation applied. New column 'ConvertedCompYearly_Log' created.")
    print("Descriptive stats for 'ConvertedCompYearly_Log':")
    print(df['ConvertedCompYearly_Log'].describe())

    # Check skewness after transformation (optional)
    transformed_skew = skew(df['ConvertedCompYearly_Log'].dropna())
    print(f"Skewness of 'ConvertedCompYearly_Log' after transformation: {transformed_skew:.2f}")

    # Visualize original vs log-transformed distribution to show effect on skewness
    plt.figure(figsize=(12, 5))
    plt.subplot(1, 2, 1)
    sns.histplot(df['ConvertedCompYearly'], kde=True, bins=30)
    plt.title('Original ConvertedCompYearly Distribution')
    plt.xlabel('Compensation')
    plt.ylabel('Frequency')

    plt.subplot(1, 2, 2)
    sns.histplot(df['ConvertedCompYearly_Log'], kde=True, bins=30, color='red')
    plt.title('Log-Transformed ConvertedCompYearly Distribution')
    plt.xlabel('Log(Compensation + 1)')
    plt.ylabel('Frequency')
    plt.tight_layout()
    plt.show()

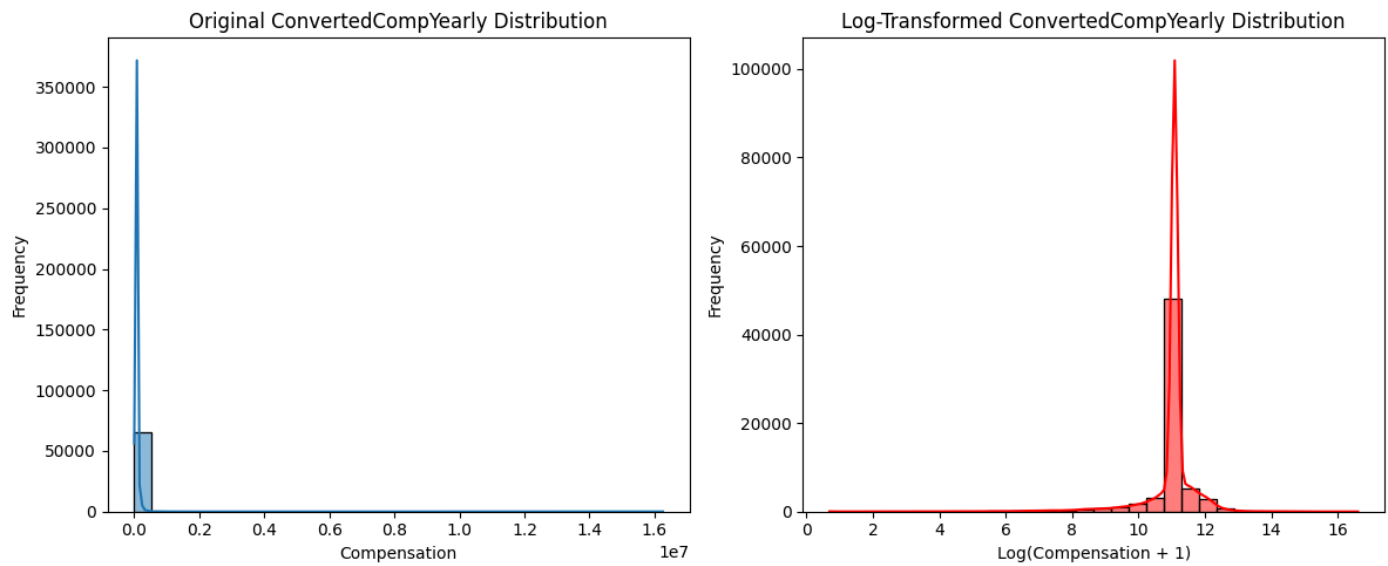
else:
    print("'ConvertedCompYearly' not suitable for log transformation (not found or not numeric).")

```

6.2 Log-transform the ConvertedCompYearly column to reduce skewness.  
Original skewness of 'ConvertedCompYearly': 87.71  
Log transformation applied. New column 'ConvertedCompYearly\_Log' created.  
Descriptive stats for 'ConvertedCompYearly\_Log':

count	65437.000000
mean	10.976053
std	0.851456
min	0.693147
25%	11.082158
50%	11.082158
75%	11.082158
max	16.604010

Name: ConvertedCompYearly\_Log, dtype: float64  
Skewness of 'ConvertedCompYearly\_Log' after transformation: -4.28



## 7. Feature Engineering

7.1 Create a new column `ExperienceLevel` based on the `YearsCodePro` column:

```
In [20]: print("\n--- 7. Feature Engineering ---")

# 7.1 Create a new column ExperienceLevel based on the YearsCodePro column:
print("\n7.1 Create a new column 'ExperienceLevel' based on the 'YearsCodePro' column.")

# Ensure 'YearsCodePro' is numeric and handle NaNs (from 5.2 or earlier)
if 'YearsCodePro' in df.columns and pd.api.types.is_numeric_dtype(df['YearsCodePro']):
    # Define conditions for different experience levels
    conditions = [
        (df['YearsCodePro'] < 3),
        (df['YearsCodePro'] >= 3) & (df['YearsCodePro'] < 10),
        (df['YearsCodePro'] >= 10) & (df['YearsCodePro'] < 20),
        (df['YearsCodePro'] >= 20)
    ]
    # Define corresponding experience level labels
    choices = ['Junior', 'Mid-level', 'Senior', 'Lead/Principal']

    # Use numpy.select to apply conditions and assign labels
    df['ExperienceLevel'] = np.select(conditions, choices, default='Unknown')
    print("New column 'ExperienceLevel' created based on 'YearsCodePro'.")
    print("\nValue counts for 'ExperienceLevel':")
    print(df['ExperienceLevel'].value_counts(dropna=False)) # dropna=False to see 'Unknown' or actu

    print("\nSample of 'YearsCodePro' and 'ExperienceLevel':")
    print(df[['YearsCodePro', 'ExperienceLevel']].head(10))

else:
    print("'YearsCodePro' column not found or is not numeric. Cannot create 'ExperienceLevel'.")

# Summary statement as per PDF (no code needed here)
print("\n--- Summary ---")
print("In this lab, you:")
print("• Explored the dataset to identify inconsistencies and missing values.")
print("• Encoded categorical variables for analysis.")
print("• Handled missing values using imputation techniques.")
print("• Normalized and transformed numerical data to prepare it for analysis.")
print("• Engineered a new feature to enhance data interpretation.")
```

## --- 7. Feature Engineering ---

7.1 Create a new column 'ExperienceLevel' based on the 'YearsCodePro' column.  
'YearsCodePro' column not found or is not numeric. Cannot create 'ExperienceLevel'.

### --- Summary ---

In this lab, you:

- Explored the dataset to identify inconsistencies and missing values.
- Encoded categorical variables for analysis.
- Handled missing values using imputation techniques.
- Normalized and transformed numerical data to prepare it for analysis.
- Engineered a new feature to enhance data interpretation.

## Summary

In this lab, you:

- Explored the dataset to identify inconsistencies and missing values.
- Encoded categorical variables for analysis.
- Handled missing values using imputation techniques.
- Normalized and transformed numerical data to prepare it for analysis.
- Engineered a new feature to enhance data interpretation.

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