Data Analytics in Software Engineering Assignment 1

Assignment: Principal Component Analysis (PCA) in Software Engineering

Dataset:

I have selected the **Al Job Market Insights** dataset for PCA assignment. This data is about the job market insights. It has 10 attributes and 500 transactions. <u>Link of Dataset</u>

Step 1: Load and Explore the Data

Load the dataset and view its basic structure to understand the attributes.

```
Job Title
                            Industry ... Remote_Friendly Job_Growth_Projection
  Cybersecurity Analyst Entertainment ...
                                                           Yes
                                                                                Growth
  Marketing Specialist Technology ...
AI Researcher Technology ...
Sales Manager Retail ...
                                                            No
                                                                               Decline
                                                           Yes
                                                                                Growth
                                                            No
                                                                                Growth
  Cybersecurity Analyst Entertainment ...
                                                                               Decline
                                                            Yes
[5 rows x 10 columns]
```

Step 2: Preprocessing

Encoding Categorical Variables: PCA requires numerical data, so we need to encode categorical variables (e.g., Job_Title, Industry, Company_Size).

Standardization: PCA also requires standardized data (mean = 0, variance = 1), so we'll scale numerical features.

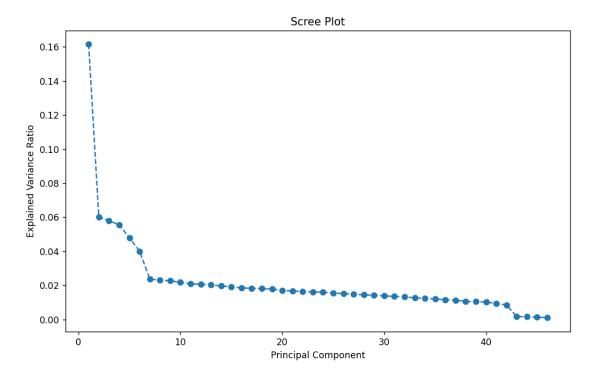
```
RangeIndex: 500 entries, 0 to 499
Data columns (total 10 columns):
 # Column
                                          Non-Null Count Dtype
0 Job_Title 500 non-null object
1 Industry 500 non-null object
2 Company_Size 500 non-null object
3 Location 500 non-null object
4 AI_Adoption_Level 500 non-null object
5 Automation_Risk 500 non-null object
6 Required_Skills 500 non-null object
7 Salary_USD 500 non-null float64
8 Remote_Friendly 500 non-null object
9 Job_Growth_Projection 500 non-null object
 9 Job_Growth_Projection 500 non-null object
dtypes: float64(1), object(9)
memory usage: 39.2+ KB
Preprocessed Data:
 num__Salary_USD ... cat__Job_Growth_Projection_Stable
             0.984671 ...
               0.125474 ...
                                                                                          0.0
               0.778561 ...
                                                                                          0.0
              0.088146 ...
                                                                                          0.0
              -0.169376 ...
                                                                                          0.0
[5 rows x 46 columns]
```

Step 3: Normalize the data (only for numeric columns)

Step 4: Apply PCA and Visualize Explained Variance

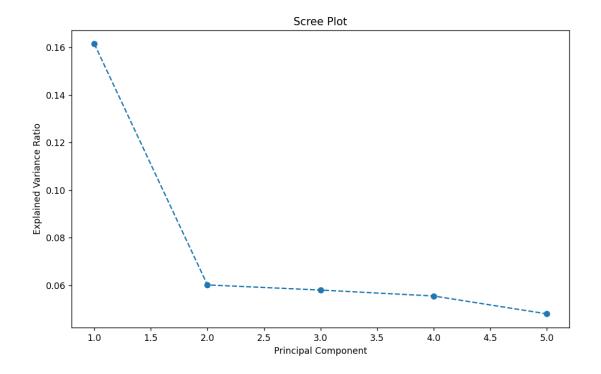
Fit PCA: We'll start by fitting PCA on the preprocessed dataset to see how many components explain most of the variance.

Scree Plot: This plot shows how much variance each component explains, helping to decide the number of components to keep.

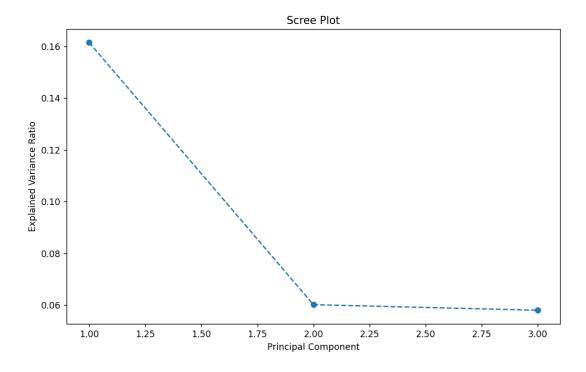


Step 5: Feature Reduction

Based on the scree plot, selected the 5 components.



Now, these results are optimal with 3 PCs.



Step 6: Analyze and Interpret Principal Components

To understand which features contribute most to each principal component, examine the loadings (the coefficients of each feature in each component).

```
Explained Variance Ratio by Principal Component:
PC1: 16.15%
PC2: 6.02%
PC3: 5.80%
Reduced Dataset Summary:
               PC1
                                                         PC4
                             PC2
                                           PC3
count 5.000000e+02 5.000000e+02 5.000000e+02
                                                5.000000e+02
    -4.973799e-17 1.065814e-17 2.486900e-17
                                                3.197442e-17
std
      1.008798e+00 6.155793e-01 6.044161e-01 5.912439e-01
min
     -2.832887e+00 -1.357219e+00 -1.147747e+00 -1.204639e+00
25%
     -6.200753e-01 -4.656863e-01 -5.645391e-01 -4.622994e-01
50%
      4.177027e-02 -4.520972e-02 -1.954146e-02 -1.544973e-02
75%
      6.315420e-01 4.173100e-01 6.004337e-01 4.587655e-01
      3.107840e+00 1.513793e+00 1.056384e+00 1.209759e+00
max
```