LinkedIn Job Trend Analysis (Web Scraping)

Objective:

To scrape job postings from LinkedIn and analyze skill demand trends across different cities and job roles, helping identify the most indemand skills by location.

Tools Used:

- Python: Main programming language
- BeautifulSoup: For HTML parsing and web scraping
- Pandas: For data manipulation and analysis
- Power BI: For Reporting and Creating Dashboards.

Steps Involved in Building the Project:

1. Web Scraping

- Used requests and BeautifulSoup to scrape job listings from LinkedIn.
- Parsed data fields like Job ID, Title, Company, Location.

2. Data Cleaning

- Handled null values in State and City columns.
- Replaced inconsistent city names (e.g., Bangalore Urban → Bengaluru).
- Removed entries with vague locations (e.g., only "India").

 Standardized job titles (e.g., removed suffixes like ", Mumbai").

3. Feature Engineering

- Created new columns: City, State, Country.
- Filled missing state data based on city mappings.

4. Exploratory Data Analysis

- Counted job titles by frequency.
- Filtered by region (Bengaluru).
- o Generated **pie charts** for top job demand in Bengaluru.

5. Insights & Recommendations

- Identified most demanded roles: Executive Assistant,
 Product Manager, etc.
- Mapped roles to career tracks (e.g., Business, Tech, HR).
- Recommended skills/tools per job type.

Conclusion:

This project successfully showcased how **job market trends** can be analyzed using real-world data. Through web scraping and data analysis, we extracted actionable insights from LinkedIn postings to support job seekers and HR professionals in understanding demand.

Key takeaways:

- Bengaluru is a major hub for roles in business, analytics, and tech.
- Standardizing data is crucial for accurate regional and job-wise analysis.
- Visualizing demand helps align career paths with market needs.