

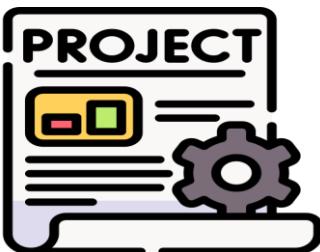
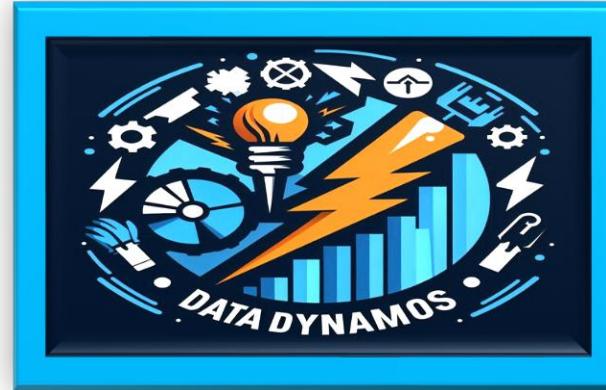


Graduation Project



رoad مصر الرقمية

Human Resources Analysis





Under Supervision of:

Dr. Sepa Mostafa



Dr. Rami Magdi



Data Analysis Project Team _DEPI

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Data Dynamos Team

Ahmed Emad Habib

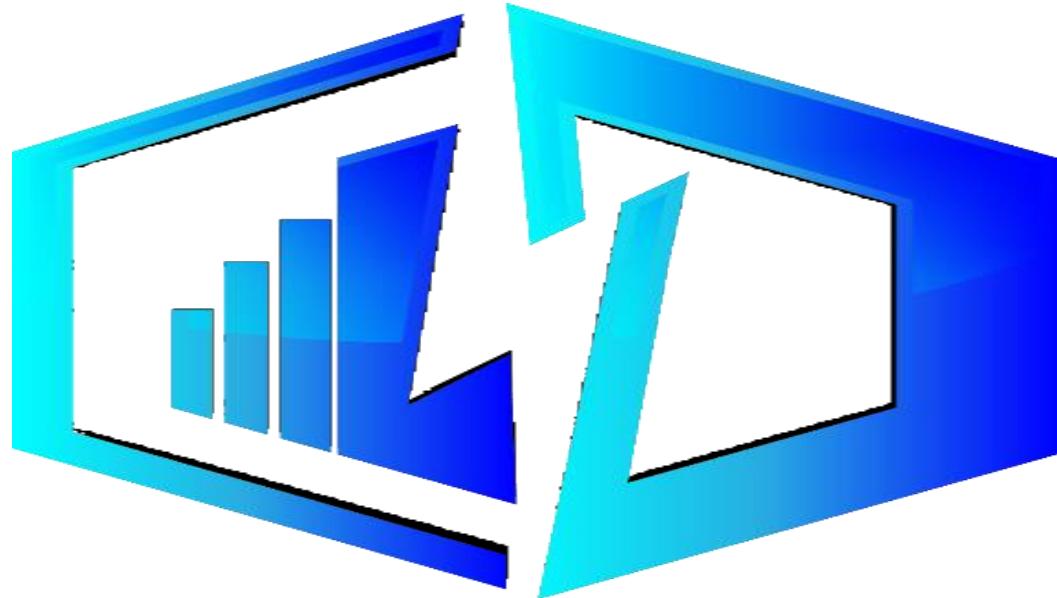
Mohammed Amgd Marzouk

Peter Rafat Adel

Mohammed Elhossiny Abrabo

Zyad Sameh Mohamed





DATA DYNAMOS

**Harnessing the Power of Data to Drive
Innovation**

About Dr. Seba



• Dr. Seba Mostafa
Data Analyst

I am Sepa, I have a BSc in Science and have been working as a senior data analyst and data analytics trainer for the past 5 years.

I have been working with both global and local companies (Udacity, Orange, Afri Labs, Carerha, ...).

- Published a paper on machine learning in an international journal.
- Trained over 1000 Students.
- Students who graduated got job offers.
- Udacity Top-Rated Instructor.
- Over 1000 hours of training done.

Meet Our Team



Ahmed Emad
Habib



Mohammed
Elhossiny Abrabo



Peter Rafat Adel

Meet Our Team



**Mohammed Amgd
Marzouk**

**Zyad Sameh
Mohamed**

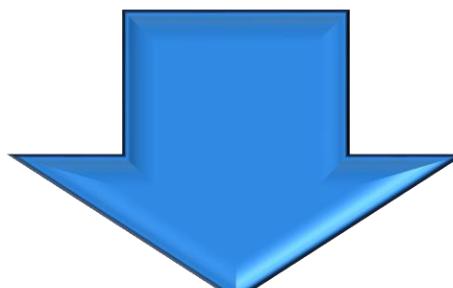
Introduction

This project focuses on analyzing the HR data of a company to gain a comprehensive understanding of employee demographics, salary trends, promotions, and work-life factors. By examining key areas such as employee turnover, salary distributions across gender and ethnicity, and correlations between education, marital status, and overtime, the project aims to provide data-driven insights. It will also explore employee-manager relationships through performance ratings and the impact of training opportunities on promotions and salaries. Various statistical methods and visual representations will be used to identify trends and patterns in the data.

Objectives

The primary objective of this project is to conduct a comprehensive analysis of HR data to uncover actionable insights that can improve employee retention, enhance job satisfaction, and optimize workforce performance. By leveraging advanced data analysis techniques and predictive modeling.

Tools: Python, Power BI, Tableau, SQL, Excel, local studio.



Analyze Age and Tenure:

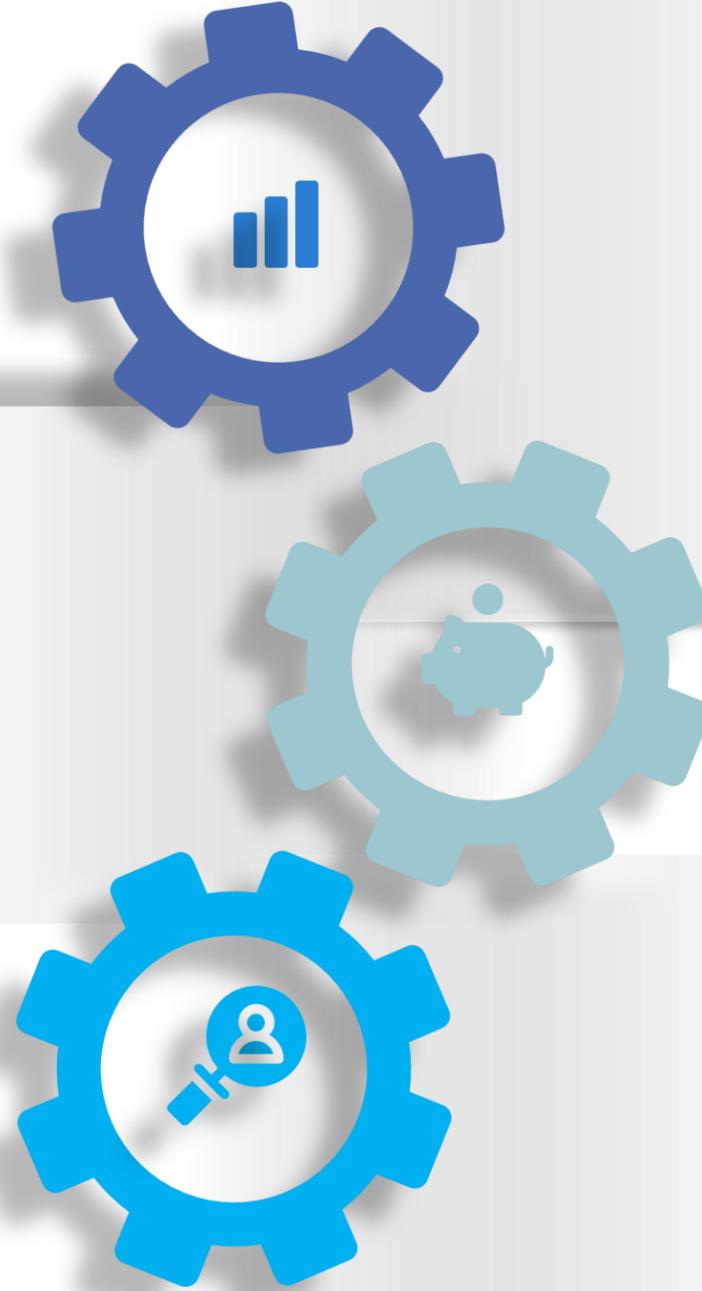
- Calculate the average, highest, and lowest employee ages.
- Identify the year with the highest employee turnover.
- Calculate the average duration employees have been with the company.
- Explore the correlation between years at the company and marital status, as well as years at the company and salary.

Examine Salary Distribution:

- Analyze salaries based on gender and ethnicity.
- Identify the highest-paid department and job role.
- Investigate the relationship between marital status and salaries, as well as distance from home and overtime.

Department and Role Analysis:

- Determine the department with the fastest promotions.
- Assess the average age of employees by state.
- Analyze the correlation between work-life balance and marital status.



Performance and Promotions:

- Examine the relationship between manager ratings and promotions.
- Calculate the difference between self-ratings and manager ratings.
- Evaluate the correlation between training opportunities taken and salaries or promotions.



Visualize Employment Trends:

- Use a line chart to display hire dates by employee count.
- Present marital status distribution with a pie chart, highlighting any correlation with salaries and overtime.

Data Collection

• Data Sources:

- Employee demographics (Gender, Age, Ethnicity, Marital Status).
- Job details (Department, DistanceFromHome, BusinessTravel, HireDate).
- Performance and tenure (PerformanceRating, YearsAtCompany, YearsSinceLastPromotion).
- Compensation (Salary, StockOptionLevel).
- Other factors (OverTime, Attrition).

The image displays three separate Excel spreadsheet windows side-by-side. The top-left window shows a mapping for 'SatisfactionID' with rows 1 through 6, where 1 is 'Very Dissatisfied', 2 is 'Dissatisfied', 3 is 'Neutral', 4 is 'Satisfied', and 5 is 'Very Satisfied'. The top-right window shows a mapping for 'EducationLevelID' with rows 1 through 5, where 1 is 'No Formal Qualifications', 2 is 'High School', 3 is 'Bachelors', 4 is 'Masters', and 5 is 'Doctorate'. The bottom window shows a large table of employee data with columns including EmployeeID, FirstName, LastName, Gender, Age, BusinessTravel, Department, DistanceFromHome, and various satisfaction and education level fields. The table contains 31 rows of data.

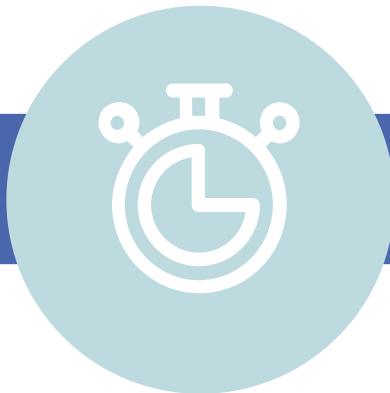
A1	SatisfactionID
1	Satisfaction
2	1 Very Dissatisfied
3	2 Dissatisfied
4	3 Neutral
5	4 Satisfied
6	5 Very Satisfied

A1	EducationLevelID
1	Education
2	1 No Formal Qualifications
3	2 High School
4	3 Bachelors
5	4 Masters
6	5 Doctorate

EmployeeID																											
A1	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y	Z		
1	EmployeeID	FirstName	LastName	Gender	Age	BusinessTravel	Department	DistanceFromHome																			
2	3012-1A41	Leontine	Simco	Female	30	Some Trav Sales	23 CA	2'	White	4 Marketing	Sales Exec Single	157718	0 Yes	1/4/2012	No	10	6	10	0								
3	CBCB-9C9 Leonerd	Aland	Male	38	Some Trav Sales	23 CA	29 CA	Asian or As	4 Marketing	HR Busine: Married	309964	1 No	1/4/2012	No	10	6	10	8									
4	95D7-1CE Ahmed	Sykes	Male	43	Some Trav Human Re	29 CA	29 CA	White	3 Computer	Engineer Married	293132	0 No	1/5/2012	No	10	10	10	0									
5	47AD-559E Ermentrud	Berrie	Non-Binar	39	Some Trav Technolog	12 IL	29 CA	White	2 Technical	Recruiter Single	49606	0 No	1/5/2012	Yes	6	1	1	6									
6	42CC-040 Stace	Savege	Female	29	Some Trav Human Re	30 NY	29 CA	Mixed or m	2 Marketing	Sales Exec Divorced	133468	1 No	1/5/2012	No	10	3	7	9									
7	C219-G2C2 Clerk	clauca Hinkins	Male	34	Some Trav Sales	45 NY	Black or Af	3 Information	Engineerin Married	259284	1 No	1/9/2012	No	10	2	6	6										
8	D906-B67 Uta	Melmar	Female	42	No Travel Technolog	3 CA	Native Hav	2 Other	Information	Sales Exec Divorced	104426	1 No	#####	No	10	3	4	6									
9	3C7D-B6E1 Joyan	Brason	Female	40	Some Trav Sales	20 IL	Black or Af	4 Marketing	Sales Exec Married	147098	1 No	#####	No	10	5	8	2										
10	3D71-BDC Alix	Blazewksi	Male	38	Some Trav Sales	4 NY	Native Hav	2 Information	Data Scien Single	69747	0 No	#####	Yes	6	5	5	1										
11	547F-CAO1 Keyley	Snoad	Female	31	Frequent T Technolog	42 CA	White	4 Computer	Data Scien Single	102022	0 No	#####	No	10	4	5	8										
12	73CF-495K Hannals	Waslin	Female	32	Some Trav Technolog	8 NY	Other	4 Information	Machine Li Single	272175	0 No	#####	No	10	7	8	2										
13	277A-6GF1 Annabel	Pablos	Female	35	Some Trav Technolog	35 NY	Asian or As	3 Marketing	Manager Single	340229	0 Yes	#####	No	10	7	9	3										
14	8BAB-B4D4 Torey	Abram	Male	38	Some Trav Sales	3 IL	White	3 Computer	Software E Divorced	48395	1 No	#####	No	10	2	2	4										
15	111D-E5E1 Edna	Alison	Non-Binar	37	Some Trav Technolog	4 NY	Mixed or m	4 Other	Senior Soft Single	97126	0 No	#####	No	10	8	10	2										
16	97F4-0B14 Verner	Powner	Male	33	Some Trav Technolog	21 IL	Black or Af	3 Information	Engineerin Married	316208	1 No	#####	No	10	8	8	7										
17	5C03-100E Willetta	Lurriman	Female	42	Some Trav Technolog	27 CA	Mixed or m	3 Marketing	Sales Exec Single	128885	0 Yes	#####	No	10	1	10	2										
18	B01B-53A Wendall	Dryden	Male	43	Some Trav Sales	34 NY	White	4 Marketing	Sales Exec Married	108315	2 No	#####	No	10	9	10	10										
19	DFA9-9901 Cale	Holston	Male	43	No Travel Sales	19 NY	Asian or As	1 Computer	Software E Married	136521	1 No	#####	No	10	3	6	1										
20	ED73-F072 Ernaline	Napoliene	Female	45	Frequent T Technolog	1 CA	White	1 Economic	Sales Exec Single	151141	0 No	#####	No	10	3	6	9										
21	C6EC-FEB3 Charlene	Severwigi	Female	38	Some Trav Sales	17 CA	White	2 Technical	ISales Exec Married	107863	1 Yes	#####	Yes	8	8	8	5										
22	D7EF-56F7 Zsa zsa	Evered	Female	39	Frequent T Sales	3 CA	Mixed or m	1 Technical	Recruiter Divorced	53616	1 Yes	#####	No	10	2	3	7										
23	C395-8C31 Curcio	Frank	Male	33	Some Trav Human Re	36 IL	White	2 Information	Software E Divorced	61298	1 Yes	#####	No	10	3	9	3										
24	E348-E12E Burnaby	Guillet	Male	36	No Travel Technolog	34 NY	Black or Af	2 Human Re	Recruiter Divorced	54132	1 Yes	#####	No	10	10	10	10										
25	B3AF-F7E5 Elvira	Ianelli	Female	45	Some Trav Human Re	36 CA	American I	4 Information	Engineerin Married	328415	0 No	#####	No	10	1	10	1										
26	469A-8121 Baxie	Rising	Male	30	Some Trav Technolog	37 CA	White	3 Computer	Machine Li Single	145337	0 No	#####	No	10	10	10	0										
27	9E22-6287 Gifford	Poynter	Non-Binar	48	Some Trav Technolog	41 CA	Asian or As	2 Marketing	Sales Exec Married	71201	2 No	#####	No	10	8	10	4										
28	D5DA-363A Rickey	Shere	Male	33	Some Trav Sales	25 NY	American I	2 Marketing	Recruiter Divorced	55682	1 No	#####	No	10	0	2	0										
29	BFF3-AAD Collen	Sedman	Female	31	No Travel Human Re	35 NY	White	2 Marketing	Sales Repr Married	63455	1 Yes	#####	No	10	4	8	7										
30	C0BE-FED Bertram	Doleman	Male	40	Some Trav Sales	21 CA	White	3 Economic	Sales Exec Single	65626	0 No	#####	No	10	2	5	1										
31	D565-2B41 Bessie	Belson	Female	47	Frequent T Sales																						

Data Wrangling

Merging Datasets:



Combined Employee,
PerformanceRating,
and EducationLevel
datasets.

Handling Nested Data:



Used data explosion
techniques to analyze
nested data.

Key Output:



Clean, merged dataset
ready for analysis.

Data Cleaning & Processing

- Handling Missing Values:

- Removed or imputed missing data.

- Correcting Errors:

- Fixed inconsistencies in data formats.

- Removing Duplicates:

- Ensured data integrity by removing duplicate records.

- Standardizing Formats:

- Normalized data for consistency.

Data cleaning

A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	
EmployeeID	FirstName	LastName	Gender	Age	BusinessTravel	Department	DistanceFromHome	State	Ethnicity	Education	EducationField	JobRole	MaritalStatus	Salary	StockOptionLevel	Overtime	HireDate	Attrition	YearsAtCompany	YearsInMostRecentRole	YearsSinceLastPromotion	YearsWithCurrManager	
1	3012-1A41	Leonelle	Simco	Female	30	Some Travel	Sales	27	IL	White	5	Marketing	Divorced	102059	1	No	1/3/2012	No	10	4	9	7	
2	CBCB-9C9D	Leonard	Aland	Male	38	Some Travel	Sales	23	CA	White	4	Marketing	Single	157718	0	Yes	1/4/2012	No	10	6	10	0	
3	95D7-1CE9	Ahmed	Sykes	Male	43	Some Travel	Human Resources	29	CA	Asian or Asian American	4	Marketing	HR Business Partner	Married	309964	1	No	1/4/2012	No	10	6	10	8
4	47A0-559E	Ermentrude	Berrie	Non-Binary	39	Some Travel	Technology	12	IL	White	3	Computer Science	Engineering Manager	Married	293132	0	No	1/5/2012	No	10	10	10	0
5	642CC-040A	Stace	Savage	Female	29	Some Travel	Human Resources	29	CA	White	2	Technical Degree	Recruiter	Single	49606	0	No	1/5/2012	Yes	6	1	1	6
6	C219-GC2E	Clerklaude	Hinkins	Male	34	Some Travel	Sales	30	NY	Mixed or multiple ethnic groups	2	Marketing	Sales Executive	Divorced	133468	1	No	1/5/2012	No	10	3	7	9
7	D906-B674	Uta	Melmar	Female	42	No Travel	Technology	45	NY	Black or African American	3	Information Systems	Engineering Manager	Married	259284	1	No	1/9/2012	No	10	2	6	6
8	3CT0-86ED	Joyan	Brason	Female	40	Some Travel	Sales	3	CA	Native Hawaiian	2	Other	Sales Executive	Divorced	104426	1	No	1/11/2012	No	10	3	4	6
9	3D71-8DC2	Alix	Blazewaki	Male	38	Some Travel	Sales	20	IL	Black or African American	4	Marketing	Sales Executive	Married	147098	1	No	1/11/2012	No	10	5	8	2
10	5476-C40D	Kayley	Snoad	Female	31	Frequent Traveller	Technology	4	NY	Native Hawaiian	2	Information Systems	Data Scientist	Single	69747	0	No	1/12/2012	Yes	6	5	5	1
11	73CF-4956	Hannis	Waslin	Female	32	Some Travel	Technology	42	CA	White	4	Computer Science	Data Scientist	Single	102022	0	No	1/12/2012	No	10	4	5	8
12	277A-A6FA	Annabella	Pablos	Female	35	Some Travel	Technology	8	NY	Other	4	Information Systems	Machine Learning Engineer	Single	272175	0	No	1/12/2012	No	10	7	8	2
13	8BAB-B4A6	Torey	Abram	Male	38	Some Travel	Sales	35	NY	Asian or Asian American	3	Marketing	Manager	Single	340229	0	Yes	1/13/2012	No	10	7	9	3
14	111D-E5EF	Edna	Alison	Non-Binary	37	Some Travel	Technology	3	IL	White	3	Computer Science	Software Engineer	Divorced	48395	1	No	1/15/2012	No	10	2	4	4
15	97F4-0814	Vernen	Powner	Male	33	Some Travel	Technology	4	NY	Mixed or multiple ethnic groups	4	Other	Senior Software Engineer	Single	97126	0	No	1/15/2012	No	10	8	10	2
16	5C03-1009	Willetta	Lurriman	Female	42	Some Travel	Technology	21	IL	Black or African American	3	Information Systems	Engineering Manager	Married	316208	1	No	1/17/2012	No	10	8	7	7
17	BD1B-53AD	Wendall	Dryden	Male	43	Some Travel	Sales	27	CA	Mixed or multiple ethnic groups	3	Marketing	Sales Executive	Single	128885	0	Yes	1/17/2012	No	10	1	10	2
18	D9A9-990E	Cale	Holston	Male	43	No Travel	Sales	34	NY	White	4	Marketing	Sales Executive	Married	108315	2	No	1/17/2012	No	10	9	10	10
19	ED73-F078	Ermaline	Napoliione	Female	45	Frequent Traveller	Technology	19	NY	Asian or Asian American	1	Computer Science	Software Engineer	Married	136521	1	No	1/18/2012	No	10	3	6	1
20	C6EC-FEB5	Charlema	Soverwright	Female	38	Some Travel	Sales	1	CA	White	1	Economics	Sales Executive	Single	151141	0	No	1/19/2012	No	10	3	6	9
21	D7EE-56FC	Zsazsa	Evered	Female	39	Frequent Traveller	Sales	17	CA	White	2	Technical Degree	Sales Executive	Married	107863	1	Yes	1/19/2012	Yes	8	8	8	5
22	C395-8C36	Curcio	Franek	Male	33	Some Travel	Human Resources	3	CA	Mixed or multiple ethnic groups	1	Technical Degree	Recruiter	Divorced	53616	1	Yes	1/19/2012	No	10	2	3	7
23	E348-E12B	Burnaby	Guillet	Male	36	No Travel	Technology	36	IL	White	2	Information Systems	Software Engineer	Divorced	61298	1	Yes	1/19/2012	No	10	3	9	3
24	B3AF-7E5B	Elvira	Iarelli	Female	45	Some Travel	Human Resources	34	NY	Black or African American	2	Human Resources	Recruiter	Divorced	54132	1	Yes	1/20/2012	No	10	10	10	10
25	469A-8121	Baxie	Rising	Male	30	Some Travel	Technology	36	CA	American Indian or Alaska Native	4	Information Systems	Engineering Manager	Married	328415	0	No	1/20/2012	No	10	1	10	1
26	9E22-6287	Gifford	Poynter	Non-Binary	48	Some Travel	Technology	37	CA	White	3	Computer Science	Machine Learning Engineer	Single	145337	0	No	1/21/2012	No	10	10	0	0
27	05DA-363A	Rickey	Shere	Male	33	Some Travel	Sales	41	CA	Asian or Asian American	2	Marketing	Sales Executive	Married	71201	2	No	1/24/2012	No	10	8	10	4
28	BFF3-AA0C	Colleen	Sedman	Female	31	No Travel	Human Resources	25	NY	American Indian or Alaska Native	2	Marketing	Recruiter	Divorced	55682	1	No	1/25/2012	No	10	0	2	0
29	30CB-EFDD	Bertram	Doleman	Male	40	Some Travel	Sales	35	NY	White	2	Marketing	Sales Representative	Married	63455	1	Yes	1/26/2012	No	10	4	8	7
30	D565-284D	Bessie	Bellson	Female	47	Frequent Traveller	Sales	21	CA	White	3	Economics	Sales Executive	Single	65626	0	No	1/27/2012	No	10	2	5	1
31	32004-D053	Joyce	Goor	Female	30	Frequent Traveller	Technology	44	CA	Black or African American	1	Computer Science	Software Engineer	Single	65808	0	No	1/28/2012	Yes	5	4	4	4
32	7FFD-C810	Claresta	Impy	Female	32	Some Travel	Technology	22	IL	White	2	Other	Data Scientist	Married	75821	1	No	1/30/2012	No	10	1	5	1
33	07B2-D67A	Rosalie	Everleigh	Male	44	Some Travel	Sales	40	NY	White	2	Economics	Manager	Divorced	285620	1	No	1/30/2012	No	10	6	8	4
34	1749-81A2	Elora	Bentjons	Female	48	Some Travel	Technology	17	CA	Black or African American	3	Information Systems	Software Engineer	Single	65860	0	No	2/1/2012	No	10	6	7	1
35	FF14-A43E	Koernraad	Namnizzi	Male	47	Some Travel	Technology	41	NY	American Indian or Alaska Native	1	Computer Science	Data Scientist	Married	40786	0	No	2/1/2012	No	10	3	5	5
36	3CD6-5587	Dorisie	Klishin	Female	31	Some Travel	Technology	20	CA	White	2	Computer Science	Software Engineer	Single	59697	0	Yes	2/2/2012	Yes	7	5	6	0
37	4FC2-A40D	Gayle	Risley	Female	32	No Travel	Technology	12	NY	Black or African American	3	Computer Science	Engineering Manager	Married	316725	1	No	2/2/2012	No	10	6	10	2
38	9AA9-DCD0	Staci	Leith	Female	28	Some Travel	Technology	14	IL	White	2	Business Studies	Software Engineer	Divorced	105984	2	Yes	2/4/2012	No	10	7	9	10
39	427A-BCG9	Pasquale	Abreheart	Male	32	Some Travel	Technology	15	CA	Mixed or multiple ethnic groups	3	Other	Analytics Manager	Married	393294	1	No	2/4/2012	No	10	2	2	2
40	A923-9E16	Alaine	Hinrichsen	Female	41	Frequent Traveller	Technology	7	CA	Black or African American	1	Business Studies	Machine Learning Engineer	Single	107008	0	No	2/5/2012	No	10	9	10	9
41	5C61-8F3A	Dyana	Gallie	Female	34	Some Travel	Technology	5	CA	White	1	Information Systems	Data Scientist	Single	127432	0	Yes	2/6/2012	No	10	8	9	9
42	CF2F-8CA3	Lindy	Rawstorne	Male	45	Some Travel	Sales	20	CA	White	2	Marketing	Sales Executive	Married	76693	1	No	2/6/2012	No	10	2	9	10
43	40A9-BE80	Joannes	McFadden	Female	39	Some Travel	Technology	31	CA	White	3	Information Systems	Software Engineer	Single	90017	0	Yes	2/7/2012	No	10	7	7	3
44	2E72-4BF1	Grace	Gohier	Male	44	Some Travel	Sales	21	CA	White	3	Marketing	Manager	Married	421642	0	No	2/7/2012	No	10	4	5	4
45	E3E9-C8F2	Jaclin	Chadbourn	Female	47	Some Travel	Technology	17	NY	White	2	Computer Science	Software Engineer	Married	46514	2	No	2/12/2012	No	10	9	9	2
46	C0C7-A3EF	Hagen	Worge	Male	48	No Travel	Sales	9	CA	White	5	Economics	Sales Executive	Single	125587	0	No	2/12/2012	No	10	10	10	4
47	79F3-8A55	Lacy	Domerc	Female	47	Some Travel	Human Resources	21	NY	Mixed or multiple ethnic groups	3	Marketing	HR Executive	Single	98608	0	No	2/14/2012	No	10	5	10	4
48	9528-A8A17	Jaquelin	Kite	Female	48	Some Travel	Technology	30	NY	Black or African American	3	Information Systems	Software Engineer	Married	72794	1	No	2/14/2012	No	10	4	5	3
49	9C71-34AE	Yvor	Walaron	Non-Binary	40	Some Travel	Technology	44	CA	White	2	Computer Science	Software Engineer	Single	69119	0	No	2/15/2012	No	10	8	10	5
50	C79A-E5E1	Alvera	Kulver	Female	32	Frequent Traveller	Technology	13	CA	White	1	Computer Science	Data Scientist	Divorced	48835	1	Yes	2/17/2012	Yes	8	5	7	0
51	5468-EE1E	Forbes	Toretta	Male	28	No Travel	Technology	44	NY	White	2	Computer Science	Senior Software Engineer	Married	218406	1	No	2/17/2012	No	10	10	10	10
52	5S0B-5A18	Paulina	Senecaut	Female	40	Frequent Traveller	Technology	27	NY	Black or African American	4	Computer Science	Software Engineer	Divorced	42333	1	No	2/19/2012	No	10	4	5	5
53	11AE-314D	Sheffie	Tonkin	Male	39	Some Travel	Sales	38	CA	Mixed or multiple ethnic groups	2	Technical Degree	Manager	Married	300299	1	Yes	2/19/2012	No	10	7	7	10
54	151D-1FF2	Kai	Lush	Female	34	Frequent Traveller	Sales	15	CA	White	2	Marketing	Sales Executive	Single	173246	0	Yes	2/20/2012	Yes	9	4	7	7
55	D584-E431	Patrica	Larryman	Female	35	Some Travel	Human Resources	22	CA	Mixed or multiple ethnic groups	1	Economics	Recruiter	Married	67264	0	No	2/22/2012	No	10	8	8	2
56	0FC8-561B	Caryl	Loving	Male	29	Frequent Traveller	Technology	14	IL	Asian or Asian American	3	Computer Science	Senior Software Engineer	Married	93591	1	No	2/22/2012	No	10	4	10	3
57	A531-8159	Carlyn	Pavese	Female	37	No Travel	Technology	16	NY	Mixed or multiple ethnic groups	3	Information Systems	Data Scientist	Divorced	82099	2	No	2/23/2012	No	10	1	7	6
58	E7A6-33A3	Amrye	Townes	Female	38	Some Travel	Technology	39	IL	White	4	Information Systems	Data Scientist	Married	83590	1	No	2/23/2012	No	10	2	3	9

Data Exploration & Transformation

- **Exploratory Data Analysis (EDA):**

- Analyzed data distribution, patterns, and relationships.

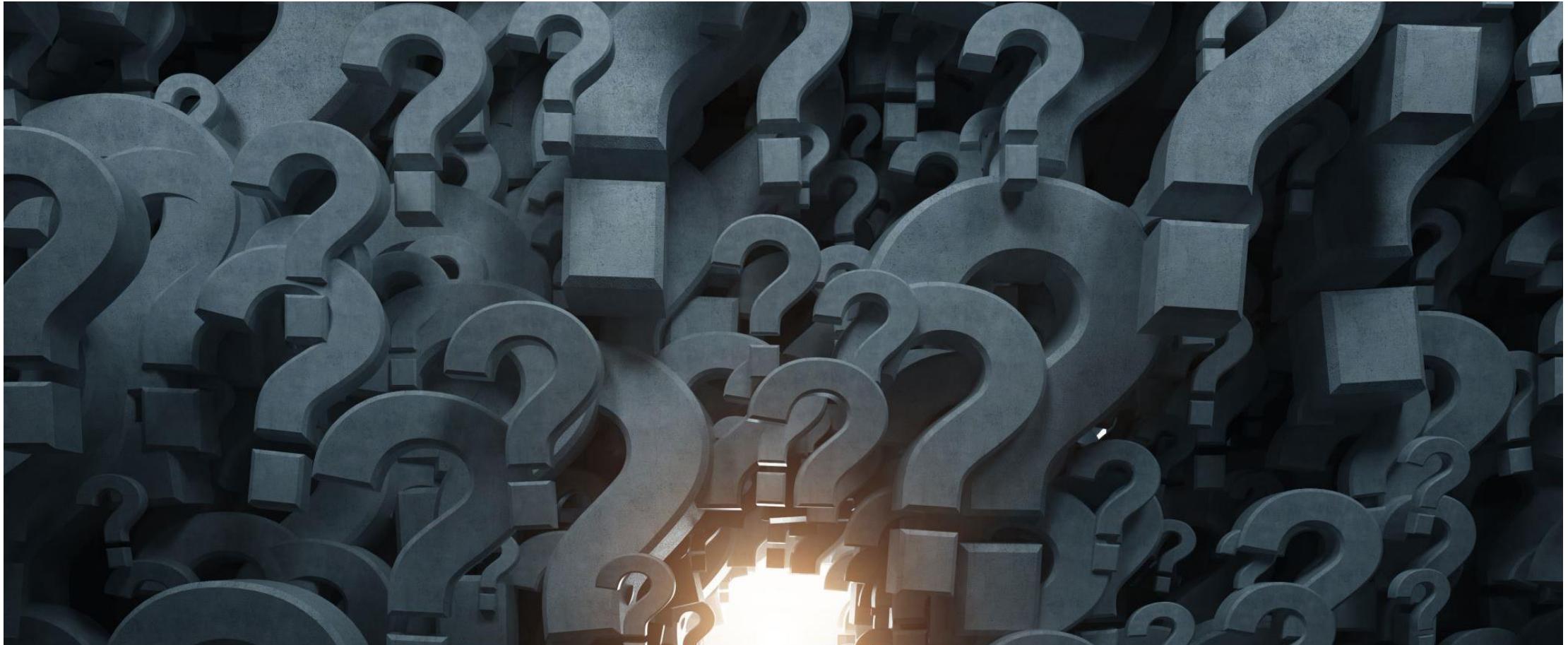
- **Data Transformation:**

- Created new variables (e.g., tenure groups, age groups).

- **Visualizations:**

- Used histograms, scatter plots, and box plots to understand data.

Key Insights Business Questions



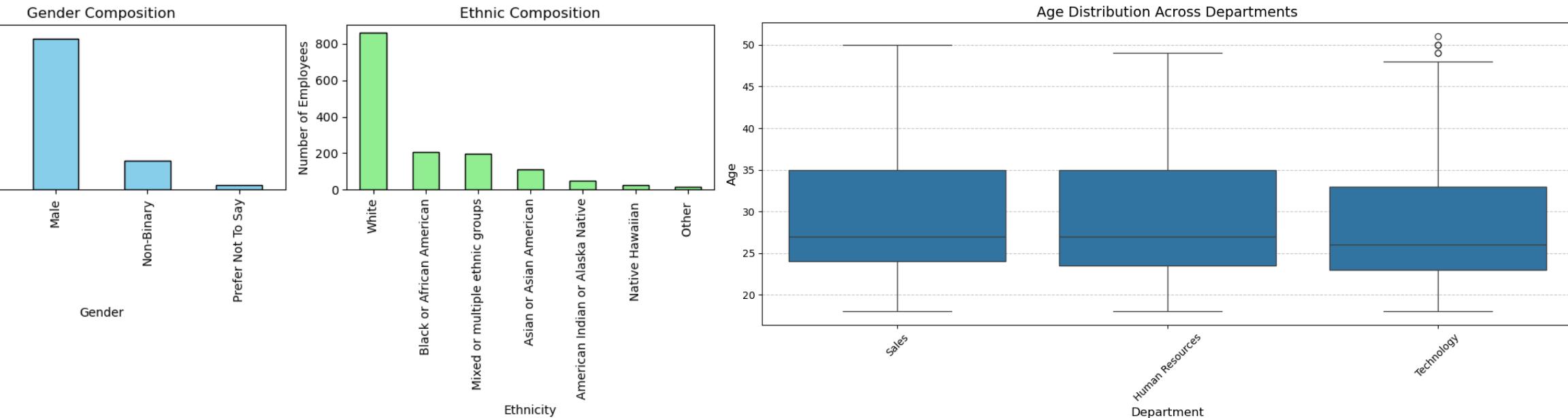
Key Insights Workforce Demographics

```
● ● ●
1 import pandas as pd
2 import matplotlib.pyplot as plt
3
4 # Correct file path
5 file_path = r'F:\data analysis\BQ\HR Data.xlsx'
6 sheet_name = 'Employee'
7
8 # Load the data
9 employee_data = pd.read_excel(file_path, sheet_name=sheet_name)
10
11 # Analyze gender and ethnic composition
12 def workforce_demographics(data):
13
14     gender_counts = data['Gender'].value_counts()
15     ethnicity_counts = data['Ethnicity'].value_counts()
16
17     # Plot gender composition
18     plt.figure(figsize=(10, 5))
19     plt.subplot(1, 2, 1)
20     gender_counts.plot(kind='bar', color='skyblue', edgecolor='black')
21     plt.title('Gender Composition')
22     plt.ylabel('Number of Employees')
23     plt.xlabel('Gender')
24
25     # Plot ethnicity composition
26     plt.subplot(1, 2, 2)
27     ethnicity_counts.plot(kind='bar', color='lightgreen', edgecolor='black')
28     plt.title('Ethnic Composition')
29     plt.ylabel('Number of Employees')
30     plt.xlabel('Ethnicity')
31
32     plt.tight_layout()
33     plt.savefig('workforce_demographics.png') # Save the plot
34     plt.show()
35
36 # Call the function
37 workforce_demographics(employee_data)
38
```

```
● ● ●
1 import pandas as pd
2 import seaborn as sns
3 import matplotlib.pyplot as plt
4
5 # Set the file path and sheet name
6 file_path = r'F:\data analysis\BQ\HR Data.xlsx'
7 sheet_name = 'Employee'
8
9 # Load the data
10 employee_data = pd.read_excel(file_path, sheet_name=sheet_name)
11
12 # Ensure the required columns exist
13 if 'Age' in employee_data.columns and 'Department' in employee_data.columns:
14     # Plot the age distribution across departments
15     plt.figure(figsize=(12, 6))
16     sns.boxplot(data=employee_data, x='Department', y='Age')
17     plt.title('Age Distribution Across Departments', fontsize=14)
18     plt.xlabel('Department', fontsize=12)
19     plt.ylabel('Age', fontsize=12)
20     plt.xticks(rotation=45)
21     plt.grid(axis='y', linestyle='--', alpha=0.7)
22     plt.tight_layout()
23     plt.show()
24 else:
25     print("Make sure the columns 'Age' and 'Department' exist in the dataset.")
26
```

Key Insights

Workforce Demographics



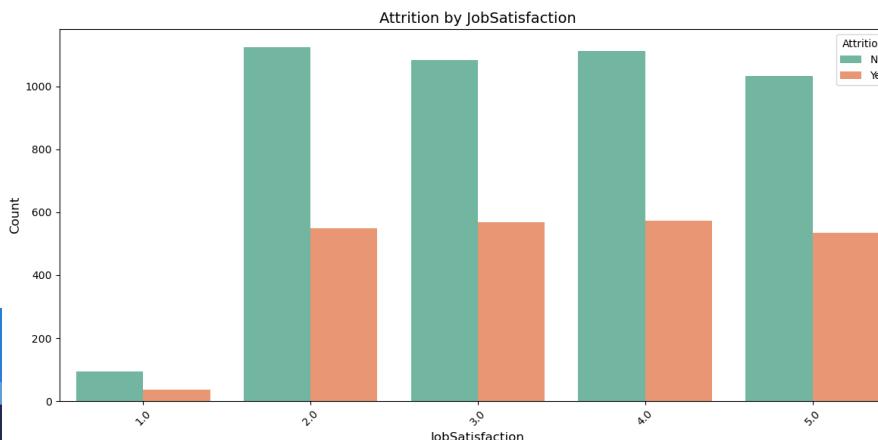
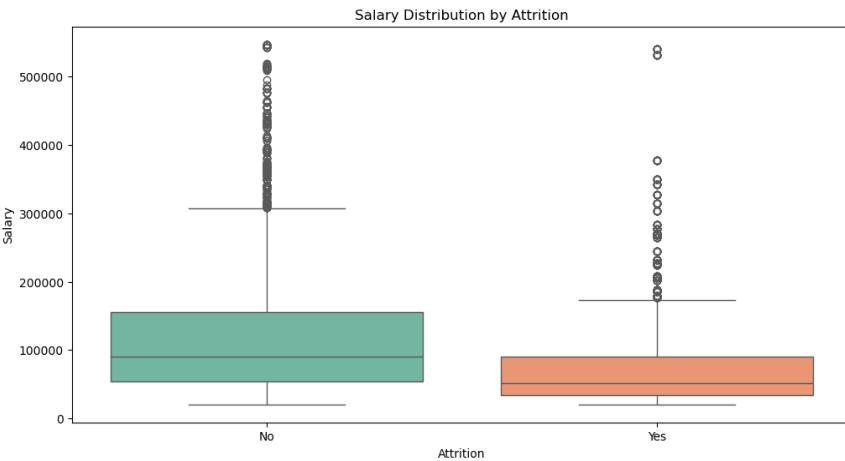
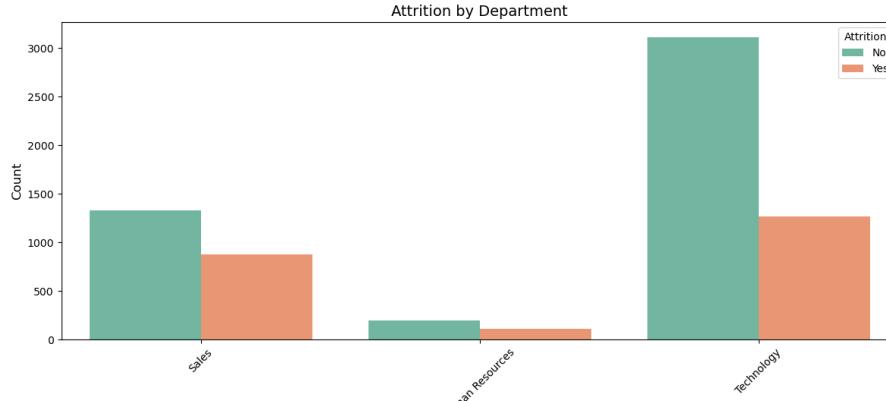
Key Insights - Employee Attrition

```

● ● ●

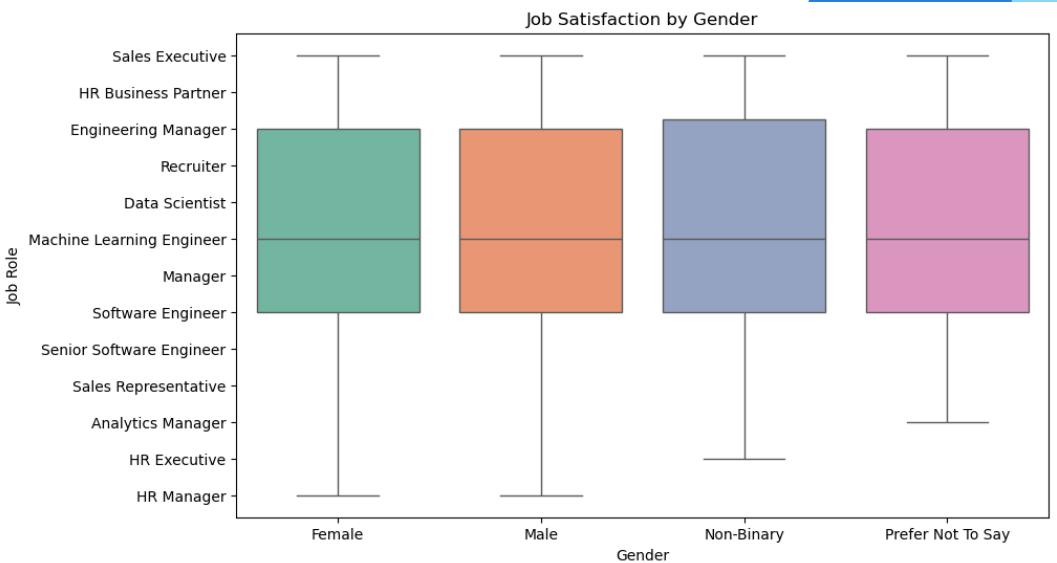
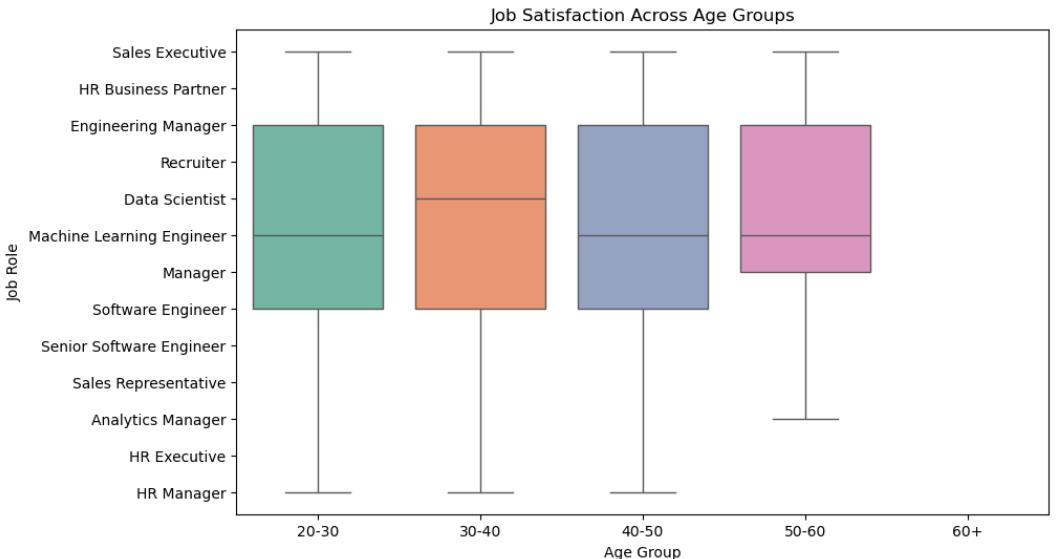
1 import pandas as pd
2 import seaborn as sns
3 import matplotlib.pyplot as plt
4
5 # Set the file path and sheet names
6 file_path = r'F:\data analysis\BQ\HR Data.xlsx'
7 employee_sheet = 'Employee'
8 performance_sheet = 'PerformanceRating'
9
10 # Load the data
11 employee_data = pd.read_excel(file_path, sheet_name=employee_sheet)
12 performance_data = pd.read_excel(file_path, sheet_name=performance_sheet)
13
14 # Merge performance data to get JobSatisfaction
15 if 'EmployeeID' in employee_data.columns and 'EmployeeID' in performance_data.columns:
16     employee_data = employee_data.merge(performance_data[['EmployeeID', 'JobSatisfaction']], on='EmployeeID', how='left')
17 else:
18     print("EmployeeID column not found in one of the datasets.")
19
20 # Rename MonthlyIncome to salary
21 if 'MonthlyIncome' in employee_data.columns:
22     employee_data.rename(columns={'MonthlyIncome': 'Salary'}, inplace=True)
23
24 # Drop rows where salary is missing
25 employee_data = employee_data.dropna(subset=['Salary'])
26
27 # Categorize salary into bins
28 bins = [0, 3000, 7000, 12000, 20000, employee_data['Salary'].max()]
29 labels = ['Low', 'Lower-Mid', 'Mid', 'Upper-Mid', 'High']
30 employee_data['salary_category'] = pd.cut(employee_data['Salary'], bins=bins, labels=labels)
31
32 # Check for the 'Attrition' column
33 if 'Attrition' in employee_data.columns:
34     # Ensure salary_category exists before analysis
35     if 'salary_category' in employee_data.columns:
36         key_factors = ['Age', 'Department', 'JobRole', 'Salary', 'JobSatisfaction']
37     else:
38         key_factors = ['Age', 'Department', 'JobRole', 'JobSatisfaction']
39
40 for factor in key_factors:
41     if factor in employee_data.columns:
42         plt.figure(figsize=(12, 6))
43         sns.countplot(data=employee_data, x=factor, hue='Attrition', palette='Set2')
44         plt.title(f'Attrition by {factor}', fontsize=14)
45         plt.xlabel(factor, fontsize=12)
46         plt.ylabel('Count', fontsize=12)
47         plt.xticks(rotation=45)
48         plt.legend(title='Attrition', loc='upper right')
49         plt.tight_layout()
50         plt.show()
51     else:
52         print(f"Column '{factor}' not found in the dataset.")
53 else:
54     print("The dataset does not contain an 'Attrition' column.")

```



Key Insights - Performance & Satisfaction

```
1 import pandas as pd
2 import matplotlib.pyplot as plt
3 import seaborn as sns
4 from scipy import stats
5
6 # Load your dataset
7 file_path = r'F:\data analysis\BQ\HR Data.xlsx'
8 sheet_name = 'Employee'
9 employee_data = pd.read_excel(file_path, sheet_name=sheet_name)
10
11 # Ensure necessary columns exist
12 required_columns = ['Attrition', 'Age', 'JobRole', 'Department', 'Gender']
13 if all(col in employee_data.columns for col in required_columns):
14
15     # Convert Attrition to numerical values (Yes=1, No=0) if not already done
16     if 'AttritionNumeric' not in employee_data.columns:
17         employee_data['AttritionNumeric'] = employee_data['Attrition'].map({'Yes': 1, 'No': 0})
18
19 # 8. Age Impact on Job Satisfaction and Attrition:
20 print("\nAge Impact on Job Satisfaction and Attrition:")
21
22 # Create age groups (e.g., 20-30, 30-40, 40-50, 50+)
23 age_bins = [20, 30, 40, 50, 60, 70]
24 age_labels = ['20-30', '30-40', '40-50', '50-60', '60+']
25 employee_data['AgeGroup'] = pd.cut(employee_data['Age'], bins=age_bins, labels=age_labels, right=False)
26
27 # 1. How does job satisfaction vary across different age groups?
28 # Assuming 'JobRole' or other columns are related to job satisfaction
29 # If you have a specific column for job satisfaction, you can use it here
30 plt.figure(figsize=(10, 6))
31 sns.boxplot(x='AgeGroup', y='JobRole', data=employee_data, palette='Set2')
32 plt.title('Job Satisfaction Across Age Groups')
33 plt.xlabel('Age Group')
34 plt.ylabel('Job Role')
35 plt.show()
36
37 # 2. What is the attrition rate for different age groups?
38 attrition_rate_by_age = employee_data.groupby('AgeGroup')['AttritionNumeric'].mean()
39 print("\nAttrition Rate by Age Group:")
40 print(attrition_rate_by_age)
41
42 # Visualize Attrition Rate by Age Group
43 plt.figure(figsize=(10, 6))
44 sns.barplot(x=attrition_rate_by_age.index, y=attrition_rate_by_age.values, palette='Set2')
45 plt.title('Attrition Rate by Age Group')
46 plt.xlabel('Age Group')
47 plt.ylabel('Attrition Rate')
48 plt.show()
49
50 else:
51     print("Some required columns are missing.")
```



Key Insights - Compensation Analysis

...
Salary Analysis Across Demographics and Departments:

Average Salary by Department:

Department

Human Resources 105804.270627

Sales 121045.081899

Technology 106396.216115

Name: Salary, dtype: float64

ANOVA Test for Salary by Department: F-statistic = 16.40, P-value = 0.0000

T-test for Salary by Gender: T-statistic = 1.61, P-value = 0.1079

Stock Option Level and Retention (Attrition):

Correlation between StockOptionLevel and Attrition: -0.17

Age Impact on Job Satisfaction and Attrition:

Average Job Satisfaction by Age Group:

AgeGroup

18-30 3.416400

31-40 3.477931

41-50 3.420327

51-60 3.000000

Name: JobSatisfaction, dtype: float64

Attrition Rate by Age Group:

AgeGroup

18-30 0.470574

...

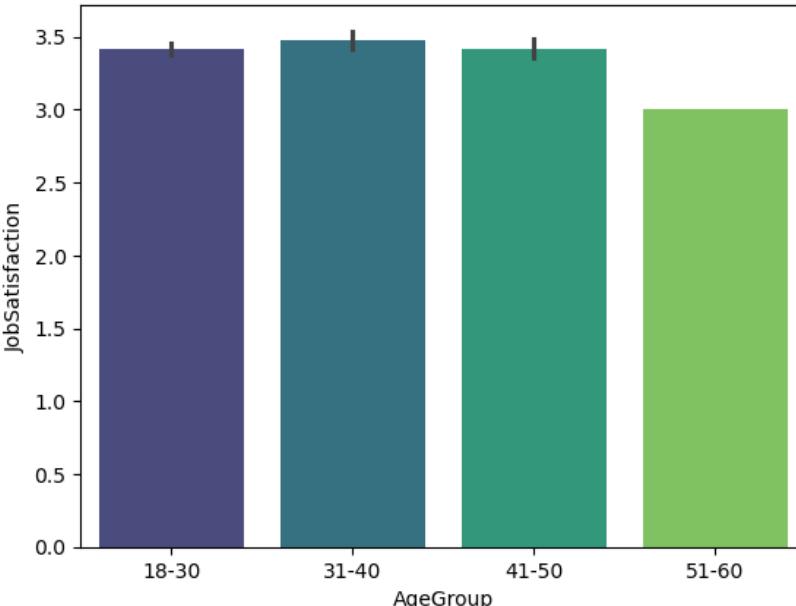
Divorced 3.405199

Married 3.440098

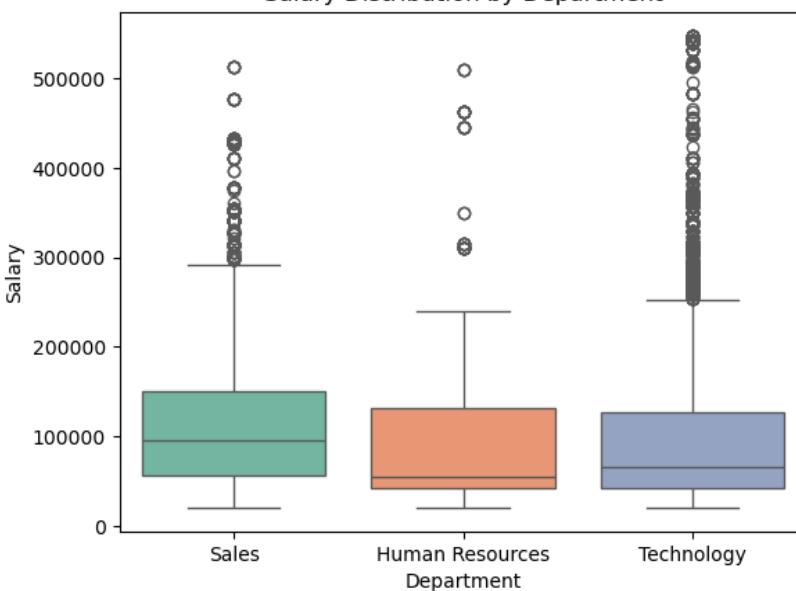
Single 3.433018

Name: JobSatisfaction, dtype: float64

Job Satisfaction by Age Group



Salary Distribution by Department

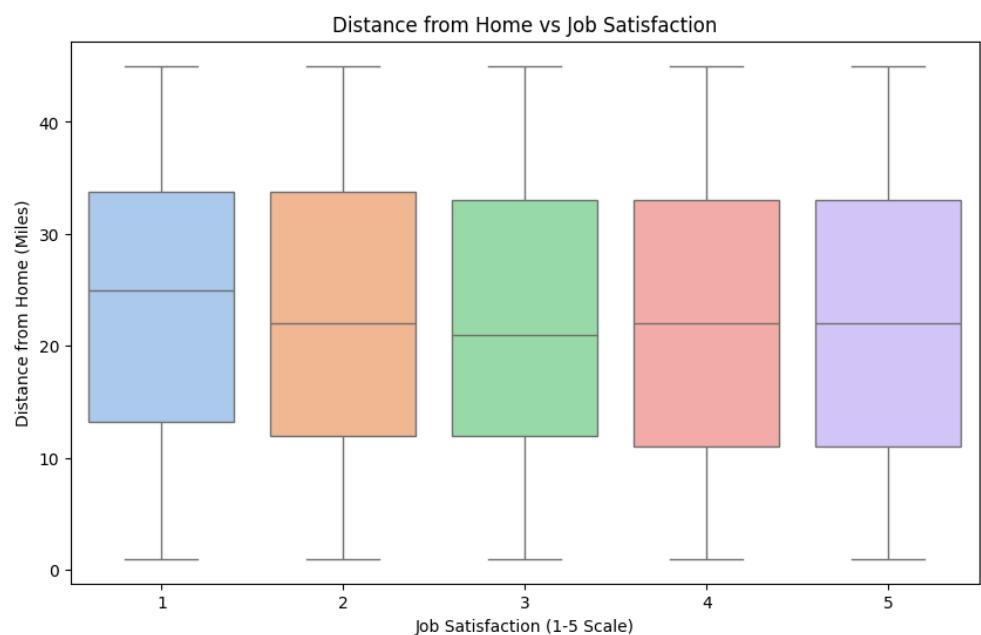
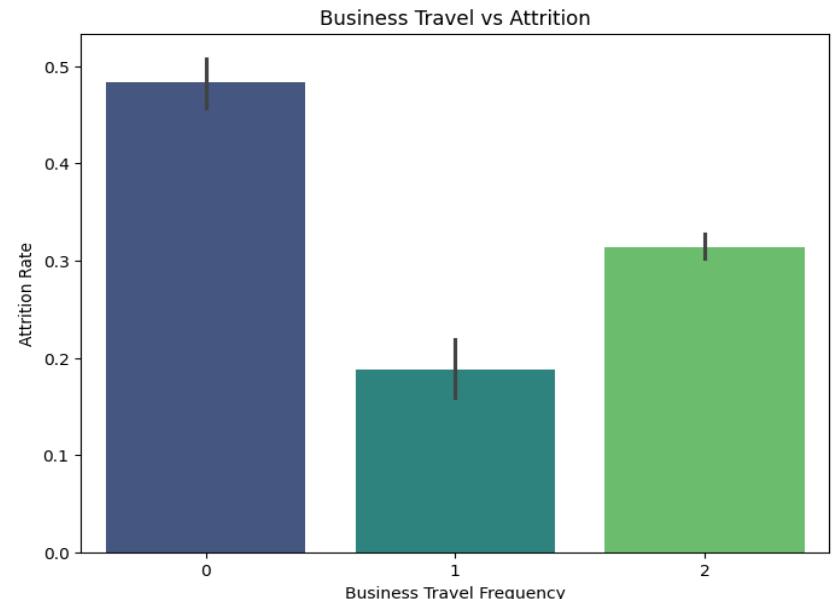


Key Insights - Geographic Insights



```
1 # Average Job Satisfaction and Attrition by Business Travel category
2 travel_satisfaction = df.groupby('BusinessTravel')[['JobSatisfaction', 'Attrition']].mean()
3 print(f"Business Travel and Job Satisfaction:\n{travel_satisfaction}")
4
5 # Visualization: Business Travel vs Job Satisfaction
6 plt.figure(figsize=(8, 6))
7 sns.barplot(x='BusinessTravel', y='JobSatisfaction', data=df, palette='viridis')
8 plt.title('Business Travel vs Job Satisfaction')
9 plt.xlabel('Business Travel Frequency')
10 plt.ylabel('Average Job Satisfaction')
11 plt.show()
12
13 # Visualization: Business Travel vs Attrition
14 plt.figure(figsize=(8, 6))
15 sns.barplot(x='BusinessTravel', y='Attrition', data=df, palette='viridis')
16 plt.title('Business Travel vs Attrition')
17 plt.xlabel('Business Travel Frequency')
18 plt.ylabel('Attrition Rate')
19 plt.show()
```

```
Business Travel and Job Satisfaction:
      JobSatisfaction  Attrition
BusinessTravel
0              3.410162   0.483063
1              3.489362   0.188216
2              3.428903   0.314346
```



Data Modeling

Predictive Analysis

Predictive Analysis



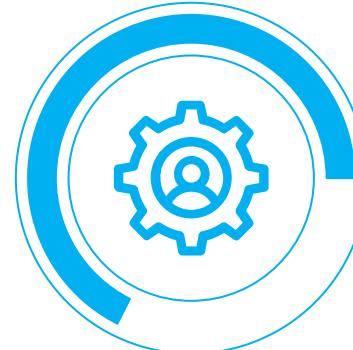
Attrition Prediction:

- The model demonstrates strong accuracy in predicting employee attrition.
- Key factors influencing attrition include overtime hours, salary levels, and the number of years spent at the company.



Promotion Readiness:

Employees with longer tenure and higher manager ratings are more likely to be promoted.



Performance Prediction:

- Salary, age, and tenure at the company are the most significant factors affecting employee performance, with salary being the most influential.



Salary Prediction:

The model estimates salaries by analyzing department, years of experience, and job role, maintaining a reasonable margin of error in its predictions.

Data Forecasting

- **Future Trends:**

- **New hires show high attrition risk without preventive measures.**
- **Overtime employees face concerning resignation trends.**
- **Hybrid work models effectively reduce turnover and boost satisfaction.**

Data Visualization

Tools: Tableau, Power BI & Excel

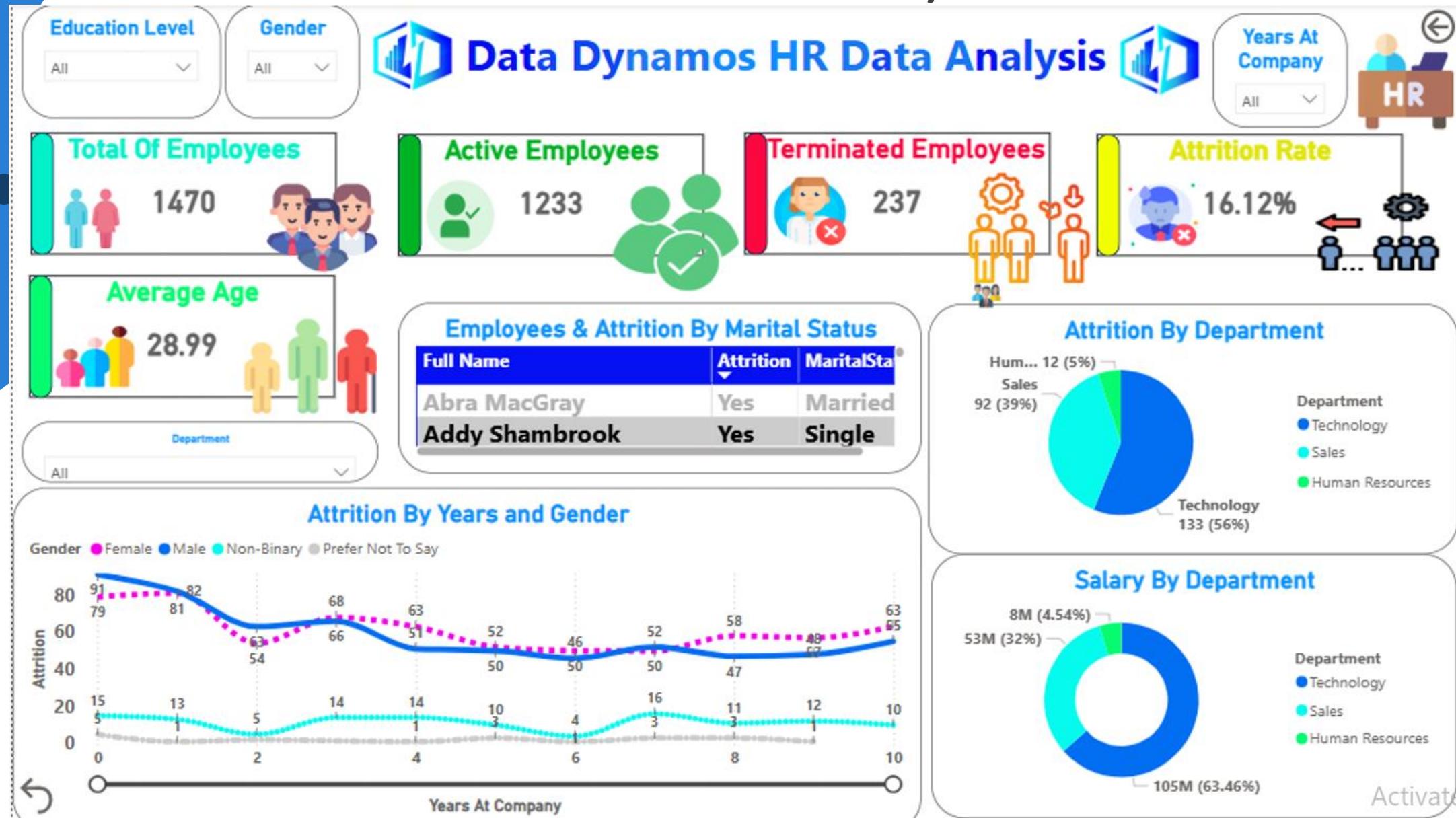
Data Visualization

Tools: Tableau, Power BI & Excel



Data Visualization

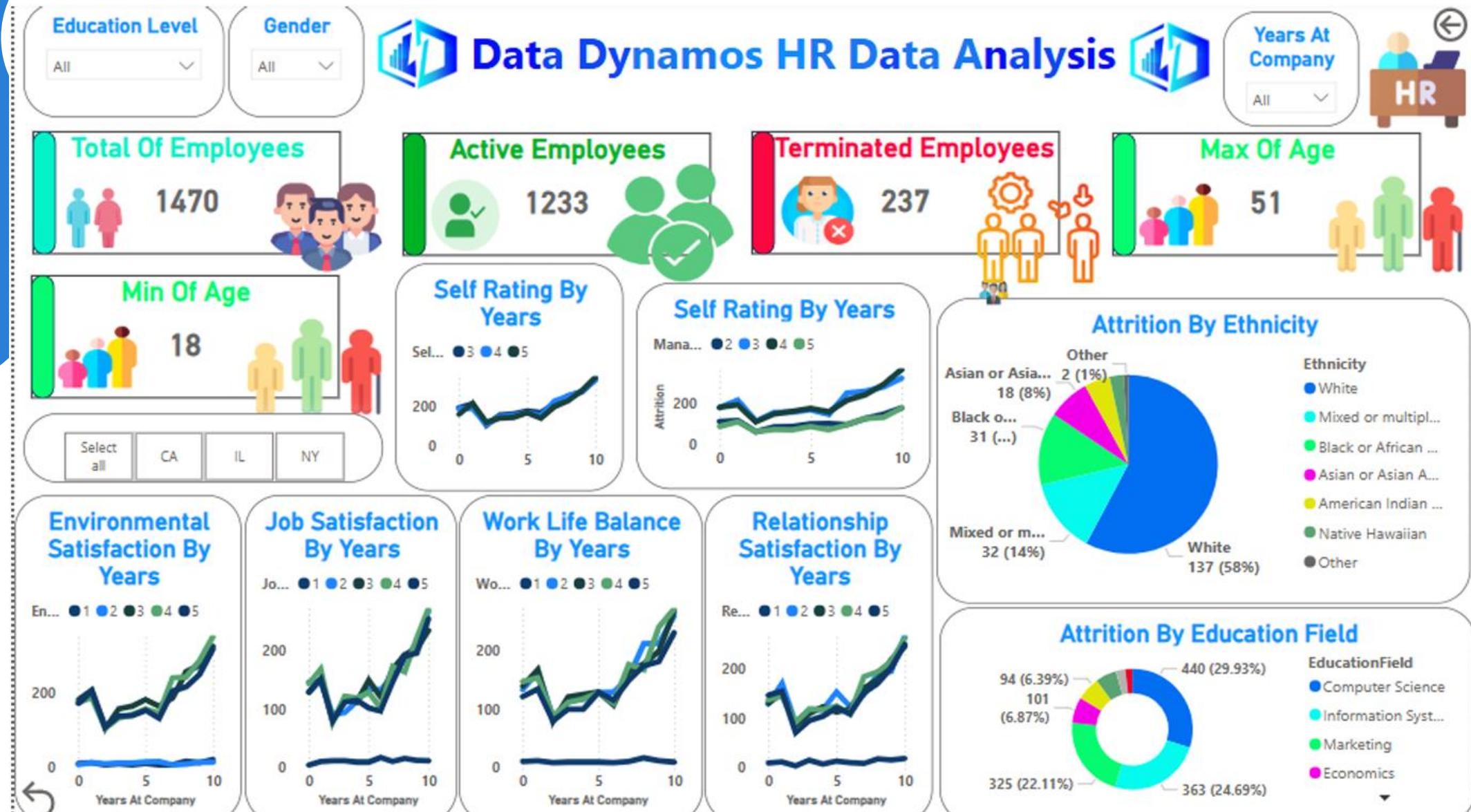
Power BI for Data Analysis



Activate

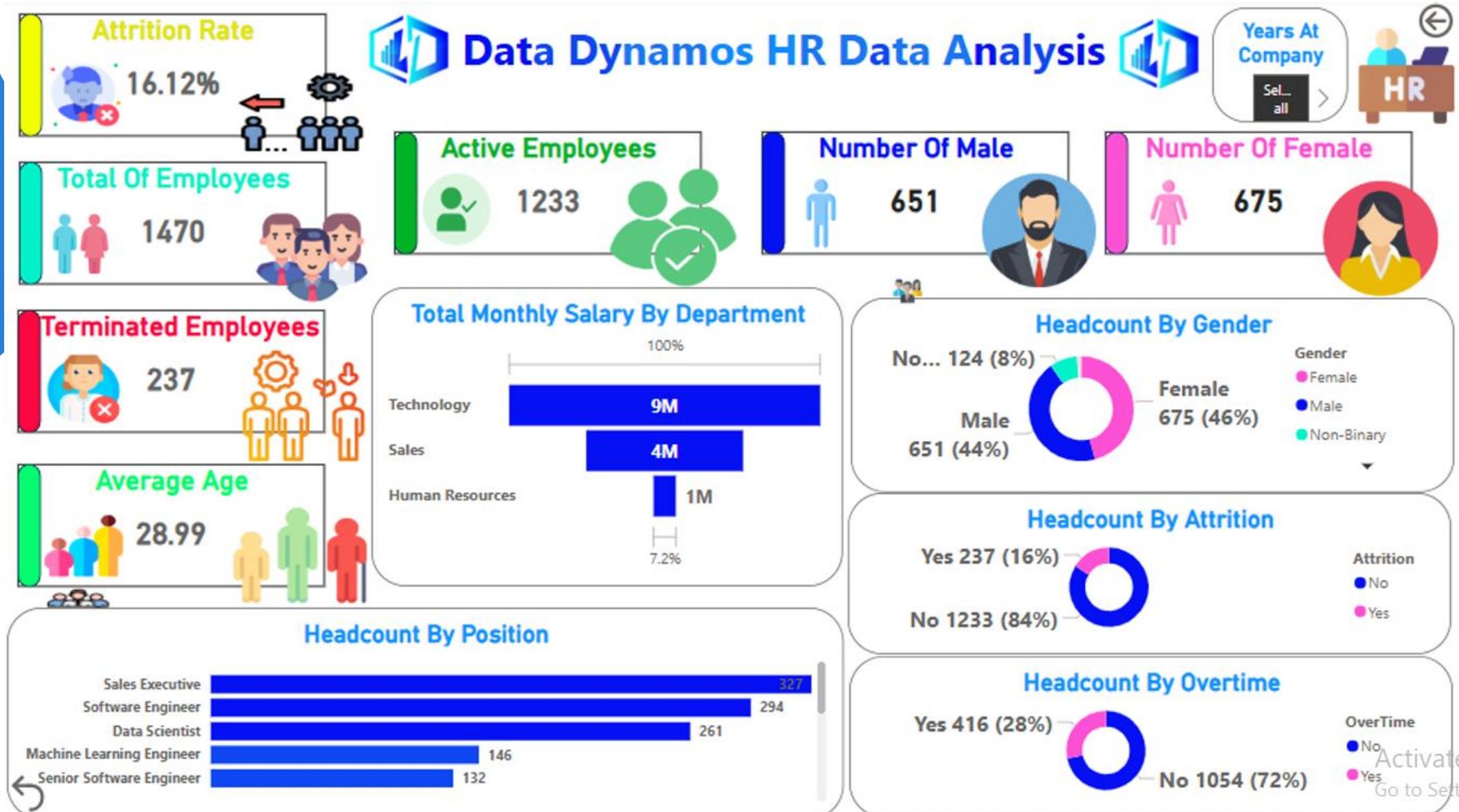
Data Visualization

Power BI for Data Analysis



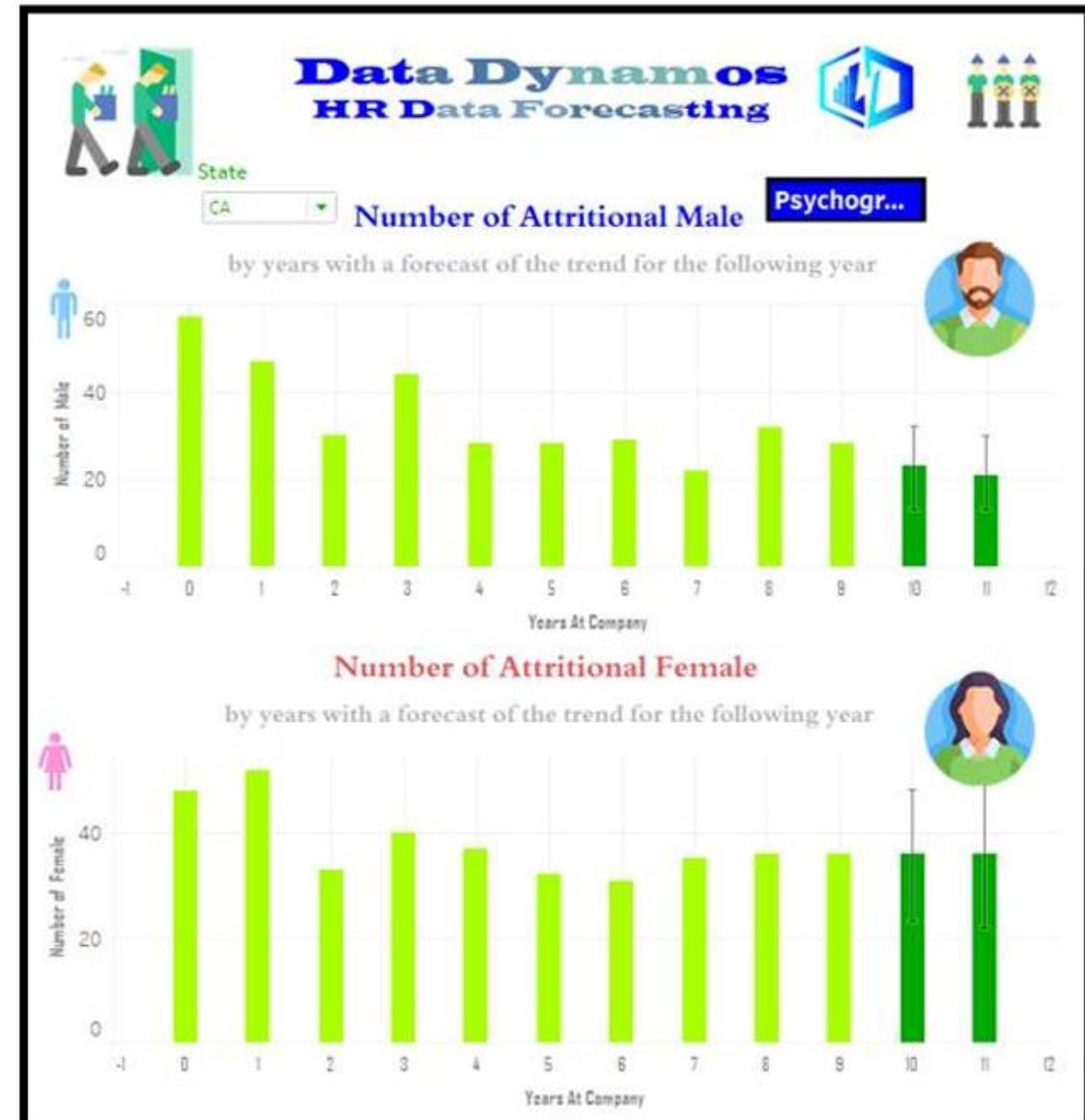
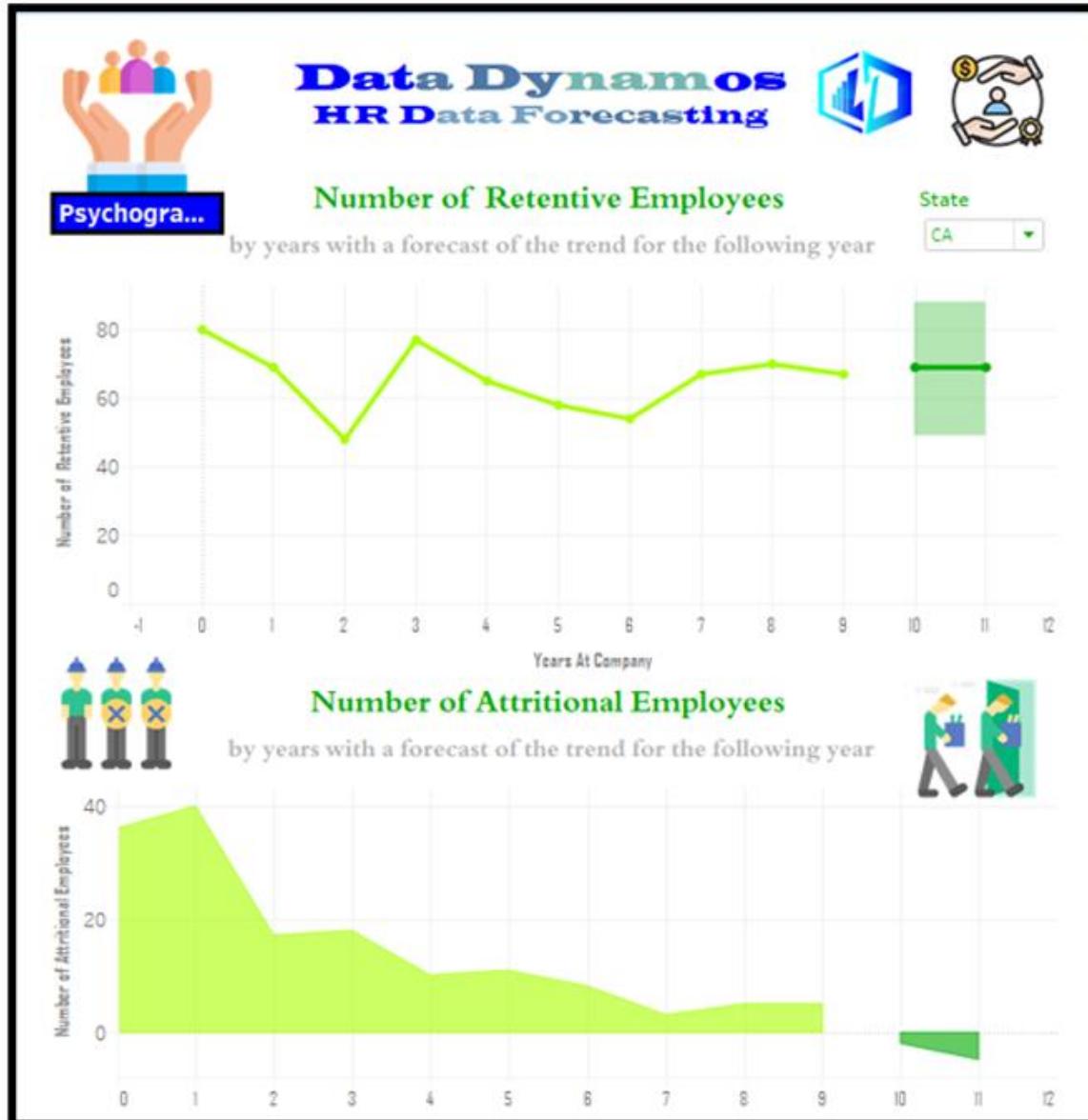
Data Visualization

Power BI for Data Analysis



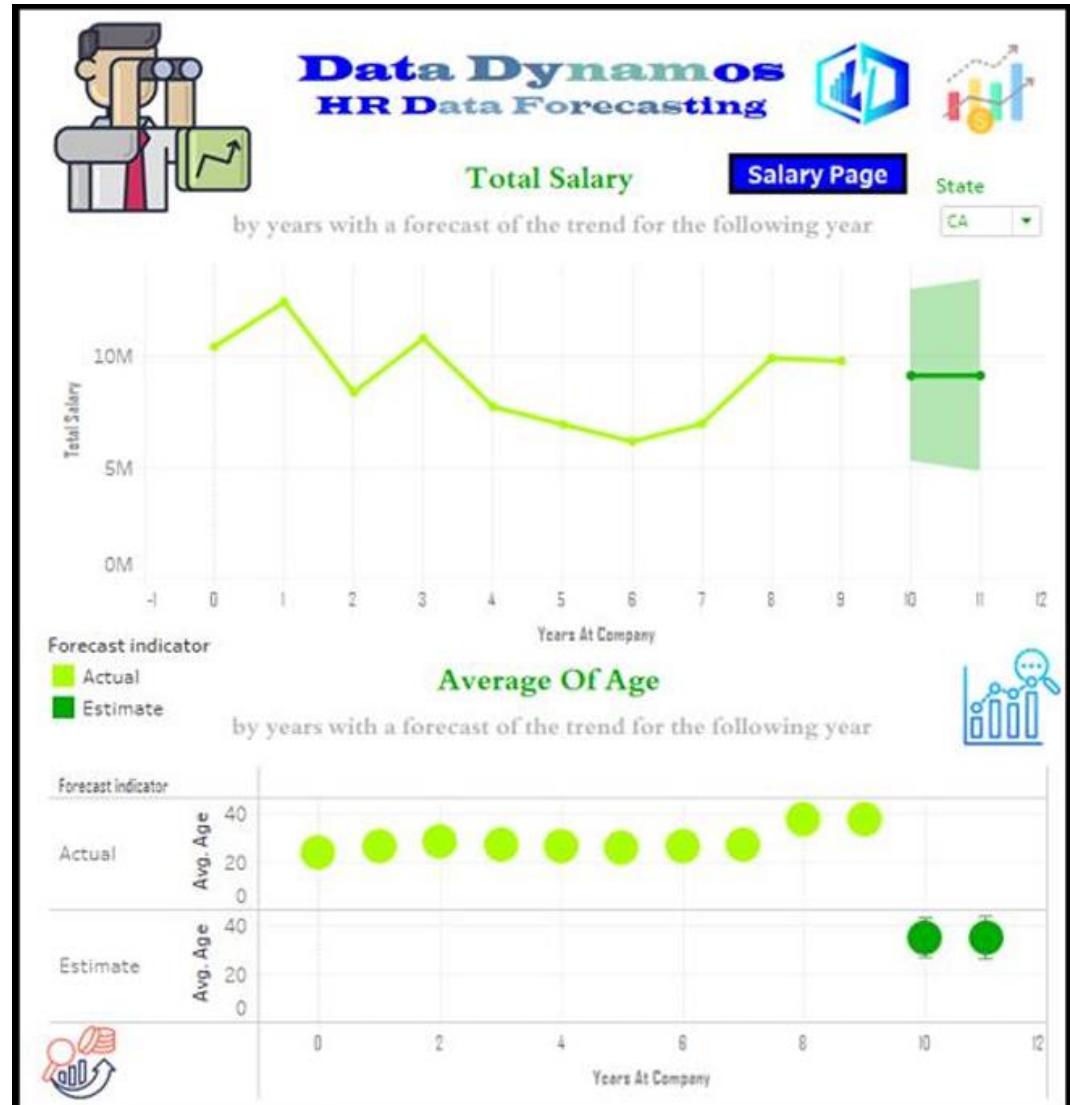
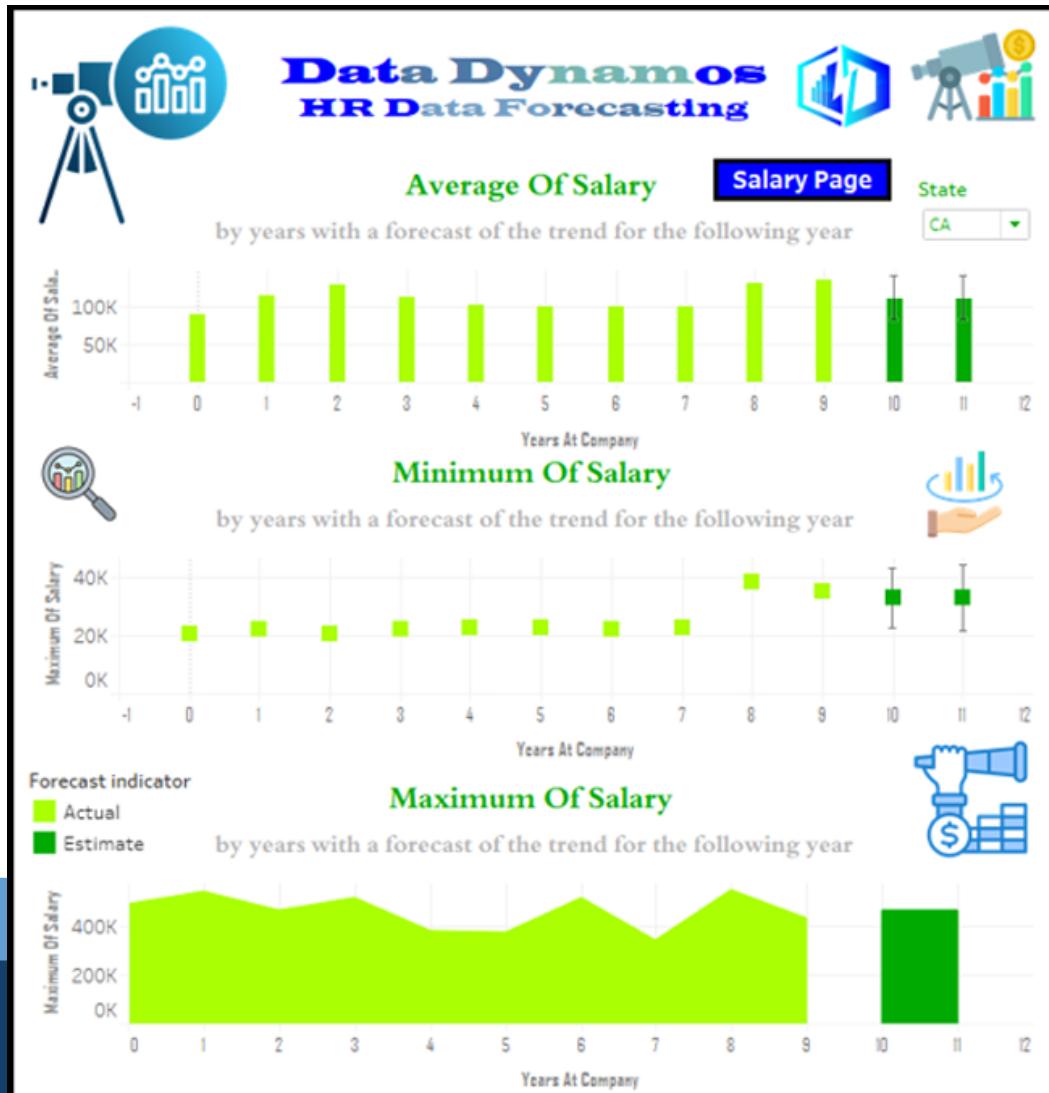
Data Visualization

Tableau for Data Forecasting



Data Visualization

Tableau for Data Forecasting



Data Visualization

excel for marketing research



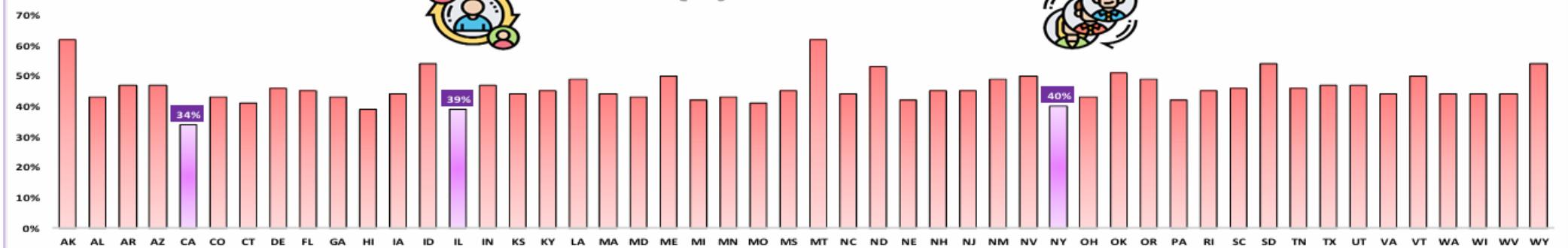
9



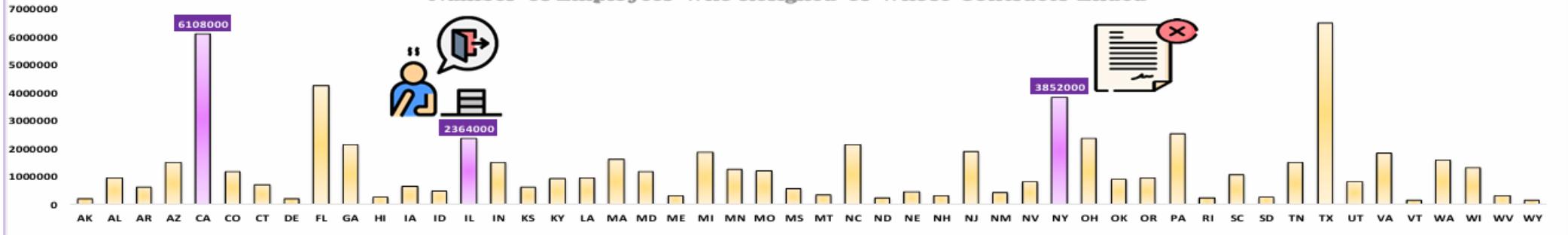
MARKET RESEARCH



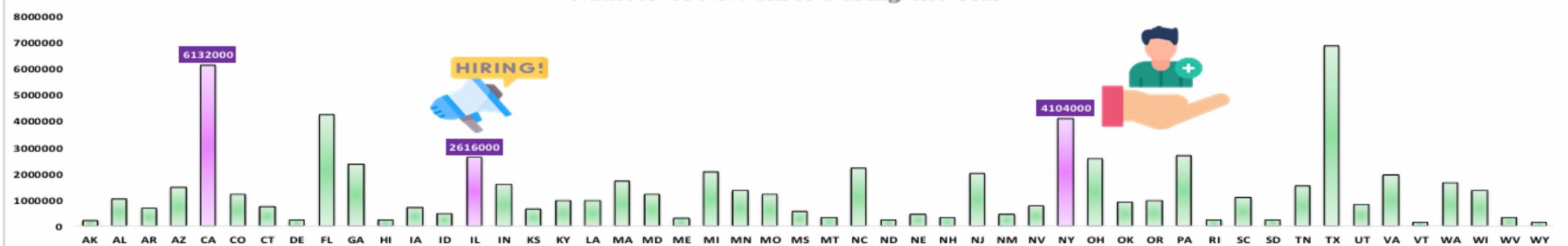
Employee Turnover Rate



Number of Employees who Resigned or whose Contracts Ended



Number of New Hires During the Year





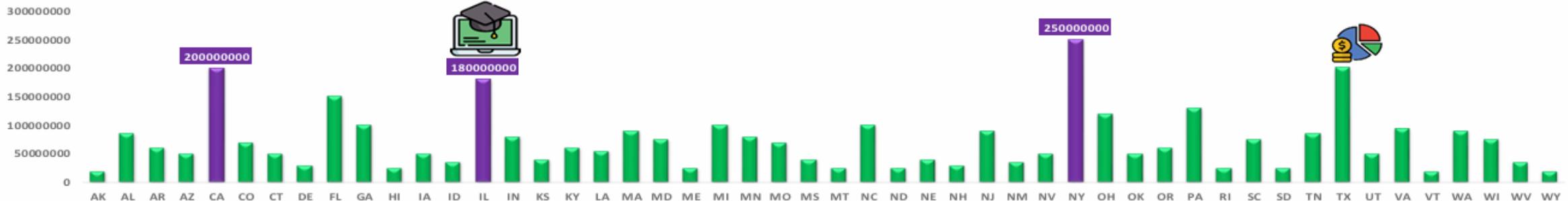
8



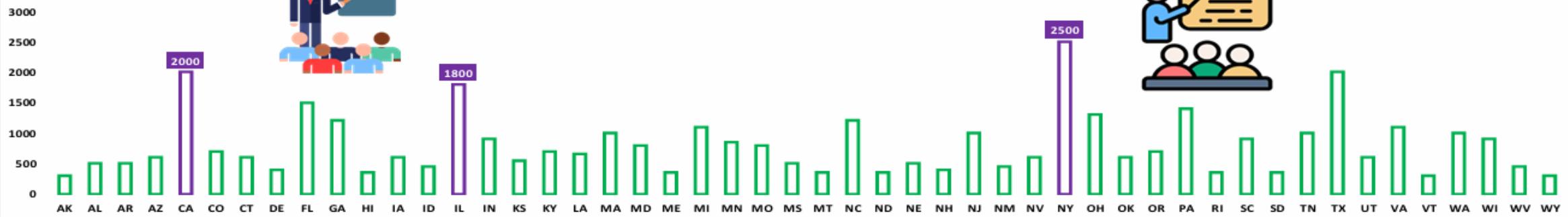
MARKET RESEARCH



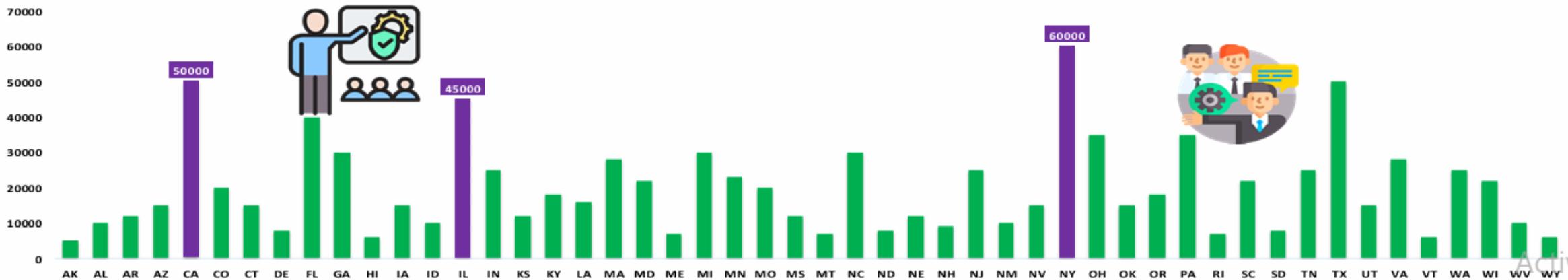
Training Budget



Number of Training Courses Conducted Annually



Number of Employees who Received Training





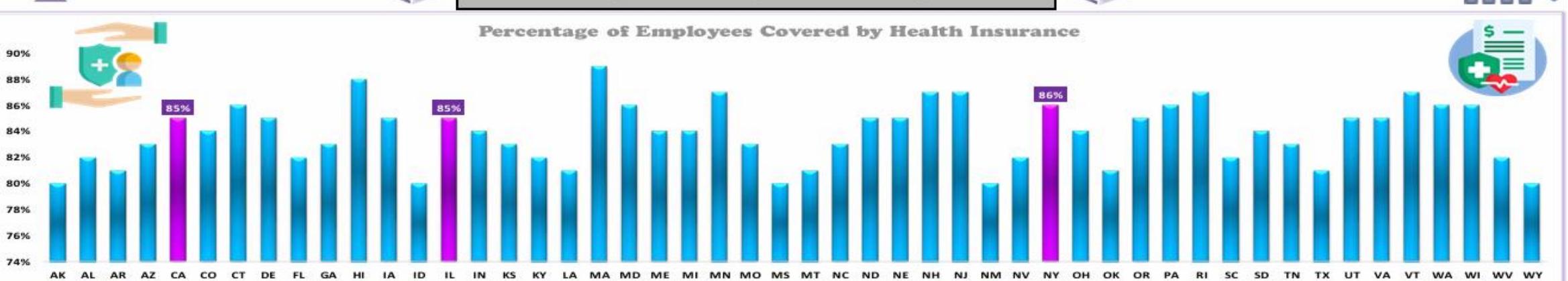
10



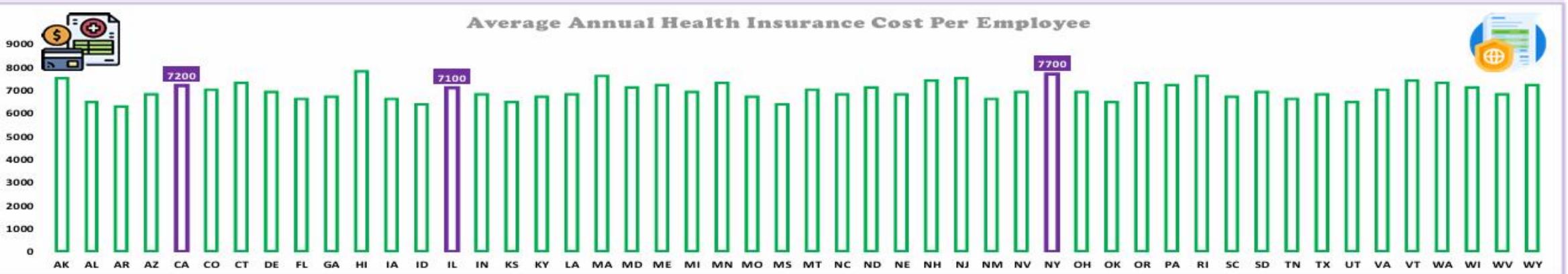
MARKET RESEARCH



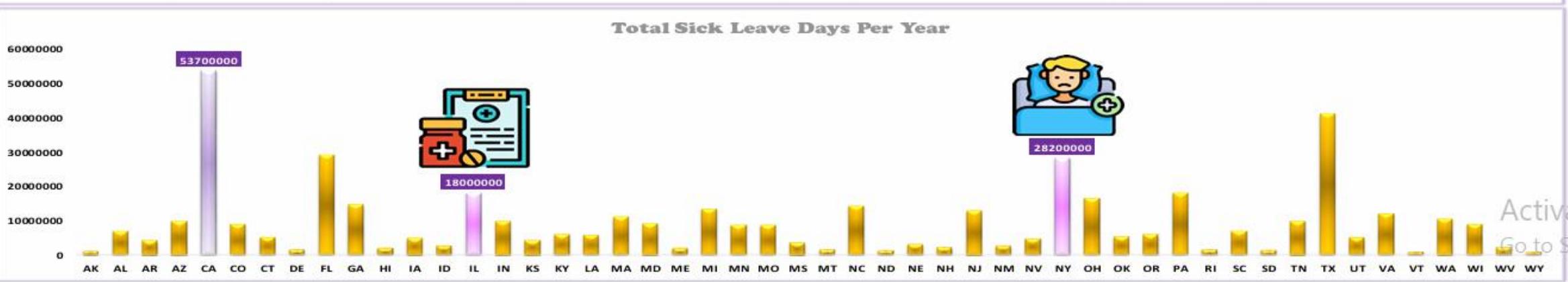
Percentage of Employees Covered by Health Insurance



Average Annual Health Insurance Cost Per Employee



Total Sick Leave Days Per Year



Data Mining

Data Mining Techniques

Clustering

1

- Grouping employees based on similar characteristics (e.g., job satisfaction, performance ratings).
- Example: Identifying high-risk groups for attrition.

- Discovering relationships between variables (e.g., Overtime → Attrition).
- Example: Employees working overtime are 3x more likely to leave.

Association Rule Mining

3



Classification

2

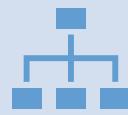
- Predicting categorical outcomes (e.g., Will an employee leave? Yes/No).
- Example: Attrition prediction using Random Forest Classifier.

- Predicting continuous outcomes (e.g., Salary, Tenure).
- Example: Salary prediction based on job role and experience.

Regression Analysis

4

Data Mining Insights



1. High-Risk Groups for Attrition:

Employees with low salaries and short tenure are at higher risk of leaving.

Employees working overtime are 3x more likely to leave.



2. Performance Trends:

High-performing employees tend to have higher salaries and longer tenure.

Employees with lower performance ratings (3/5) have the highest resignation rates.



3. Training Effectiveness:

Employees who participate in training programs show improved performance and higher promotion rates.

Training participation is positively correlated with job satisfaction.

Data-Driven Decision Making

• Recommendations:

- Improve work-life balance and reduce overtime.
- Enhance compensation and benefits.
- Provide clear career paths and training opportunities.
- Strengthen diversity and inclusion initiatives.

Data Storage

Platforms Used:

- GitHub and Google Drive.

Repository Structure:

- Organized by project stages (Data Collection, Data Cleaning, etc.).

Accessibility:

- Secure and easily accessible for team members.

Market Research Analysis

Labor Force

- Key states demonstrate strong workforce participation, indicating robust talent availability for strategic hiring initiatives. (like: CA, IL, NY)

Unemployment Rate

- The current tight labor market creates competitive hiring conditions, requiring careful consideration of timing, compensation, and training investments.

Average Annual Salaries

- Significant regional salary differences emerge, highlighting the importance of market-aligned compensation strategies for talent retention and productivity.

Employment Type

- The mix of employment arrangements provides valuable insights for optimizing benefits and workforce structure to balance organizational flexibility with operational stability.

Salary Structure

- Disparities in minimum and maximum pay reveal equity and skill demand.
- Important for compensation strategy and budget planning.

Performance Evaluation

- Supports promotion planning, training needs, and employee engagement.

Insurance & Benefits

- Comprehensive benefits and insurance coverage serve as a competitive advantage in both attracting top talent and improving employee retention.

Total Employees

- Reflects business expansion or downsizing.
- Aids in resource planning and performance analysis

Training & Development

- Significant organizational investment in continuous learning programs demonstrates commitment to workforce capability building and future readiness.

Turnover & Recruitment

- Employee retention patterns directly influence recruitment strategies, particularly during periods of business growth or workforce instability.

Absenteeism & Leave

- Indicates job satisfaction and organizational health.
- Helps optimize workforce planning.

PESTEL Analysis

Economic

- Significant variations in compensation levels highlight the need for equitable pay structures.
- External economic factors including inflation and currency fluctuations may pressure salary budgets, while overall economic conditions directly shape talent acquisition strategies.

P

E

S

Political

- Government policies on labor laws, wages, insurance, and leave directly impact hiring and compensation.
- Political stability and tax policies influence operational costs.
- Union influence and potential strikes may affect workforce stability.

Social

- Strong employee benefits and work-life balance provisions contribute to workforce satisfaction.
- Rising expectations for diversity, flexible work, and career development.
- Sick leave policies vary by location and are loosely regulated.

Environmental

- Remote work can reduce environmental impact from commuting.
- Opportunities exist for CSR and eco-friendly initiatives.
- New environmental regulations may require office policy updates.

T

E

L

Technological

- Substantial investments in digital upskilling programs position the workforce for future readiness.
- Tools like Power BI and Tableau enhance decision-making.
- Automation and cybersecurity are key areas for improvement.

Legal

- High health insurance coverage shows compliance.
- Adherence to labor laws, data privacy, and anti-discrimination regulations is essential.
- Regular contract reviews and legal audits are recommended.

Strengths

Weaknesses

Opportunities

Threats

SWOT Analysis



STRENGTHS

- Competitive salaries and long employee tenure.
- High job satisfaction and productivity.



WEAKNESSES

- High turnover and weak incentive programs.
- Salary disparities and lack of training opportunities..



OPPORTUNITIES

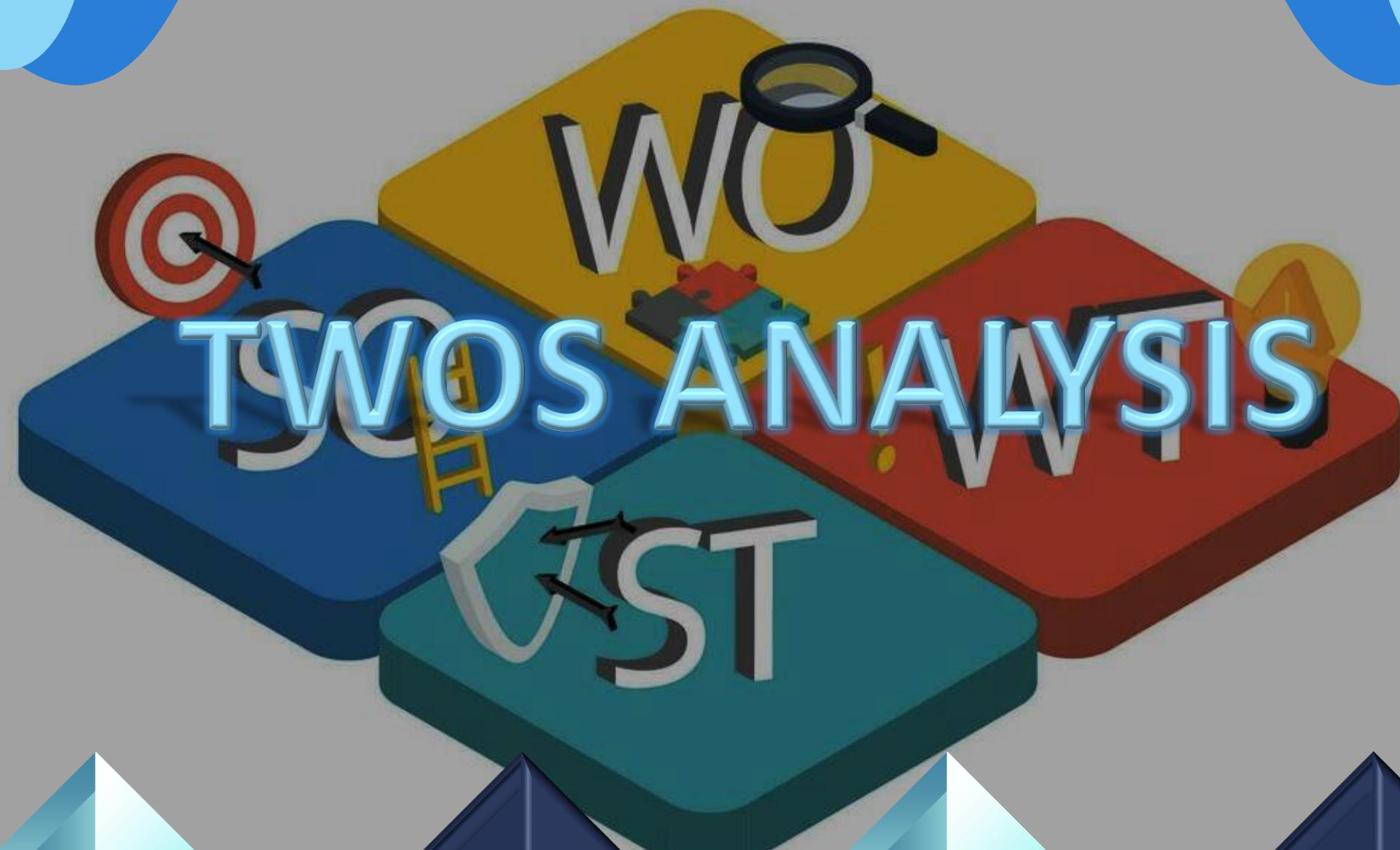
- Improve training programs and offer remote work.
- Leverage technology and expand into new markets.



THREATS

- Intense competition for talent and rising recruitment costs.
- Work pressures and new regulations





SO

ST

WO

WT

TWOS ANALYSIS

SO

1. Develop training programs to enhance skills based on evaluations.
2. Leverage competitive salaries to attract and retain talent.
3. Increase travel and job exchange opportunities to improve employee efficiency.



ST

WO

WT

TWOS ANALYSIS

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ST

1. Enhance job stability through long-term incentive policies.
2. Develop an attractive work environment to reduce external competition for skilled employees.
3. Improve promotion policies to ensure high-performing employees stay.



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WO

1. Improve financial incentives and rewards to create a more competitive work environment.
2. Implement sustainable professional development plans to empower employees.
3. Enhance work-life balance through flexible policies.



WT

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WO

1. Improve financial incentives and rewards to create a more competitive work environment.
2. Implement sustainable professional development plans to empower employees.
3. Enhance work-life balance through flexible policies.



WT

1. Reduce attrition rates through work environment improvements and stress reduction initiatives.
2. Review recruitment and retention policies to ensure workforce stability.
3. Introduce psychological support and career counseling programs to reduce job stress and increase productivity.





Strengths, Opportunities, Aspirations, and Results

SOAR ANALYSIS

Aspirations

Results

Strengths

Opportunities



Strengths, Opportunities,

Aspirations, and Results

SOAR ANALYSIS

Strengths

- **High Workforce Satisfaction:** Strong employee approval of workplace relationships and organizational culture.
- **Sustainable Work-Life Integration:** Established policies supporting employee well-being and balance.
- **Talent Retention Edge:** Competitive compensation packages and lower turnover rates than industry peers.

SOAR ANALYSIS

Opportunities

- Enhance training programs and attract young talent.
- Use performance analytics and expand job benefits.

Strengths

SOAR ANALYSIS

Aspirations

- Become a regional leader in flexible work innovation
- Prioritize internal talent growth and decision-making

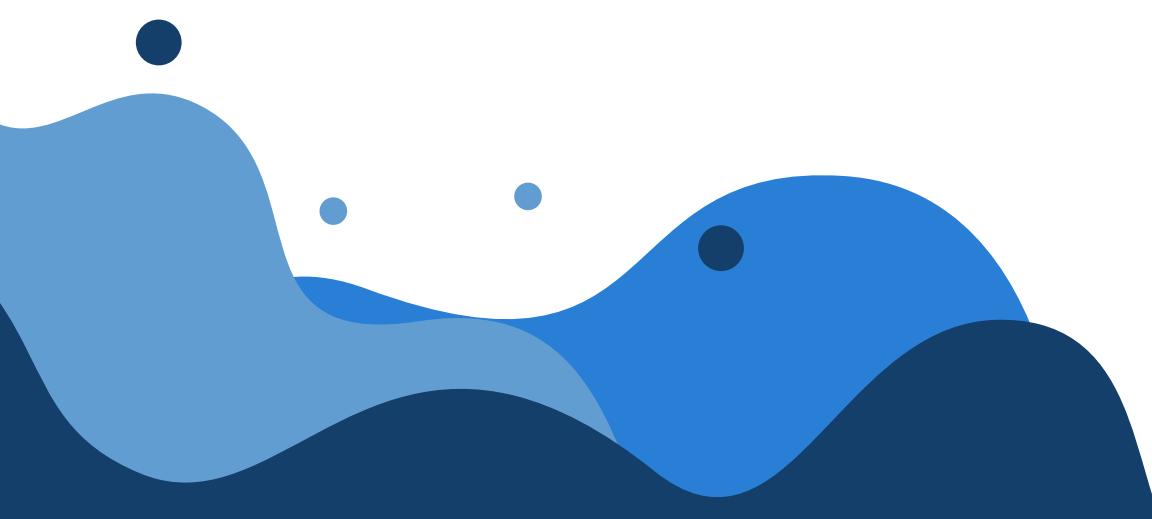
SOAR ANALYSIS

Results

- Improved employee retention and skill development
- Streamlined HR processes through tech adoption

Strengths

VRIO Analysis



1) Value:

High-performing employees, strong training programs, and job satisfaction contribute to valuable human capital.

2) Rarity:

Specialized expertise, multilingual talent, and technical proficiency create differentiation.

3) Imitability:

- Internal training programs and learning culture are hard to replicate.
- Significant investment in employee development increases uniqueness.

4) Organization:

Stable teams, effective HR strategies, and low attrition ensure alignment with business goals.

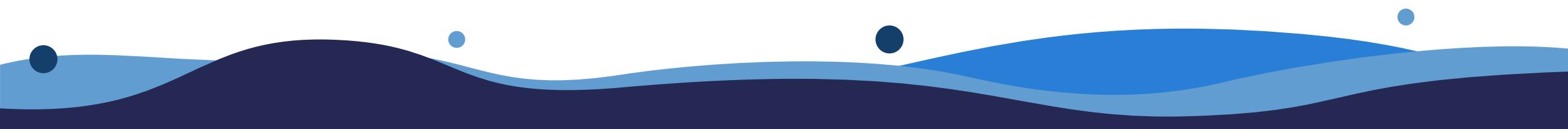
Conclusion

•Summary of Insights:

- The company has valuable and rare human resources with strong training programs.
- High turnover and work-life balance issues need addressing.
- Predictive models and strategic recommendations can improve retention and performance.

•Next Steps:

- Implement recommended strategies to improve retention and performance.
- Continuously monitor workforce trends and adjust policies as needed.





Any Question?

THANKS!