

```
In [1]: # import libraries
import pandas as pd
import numpy as np
```

```
In [2]: # Load the hr_dataset cleaned
df= pd.read_csv('hr_cleaned_dataset.csv')
df.head()
```

Out[2]:

	EmployeeID	Age	Department	SatisfactionScore	LastEvaluationScore	NumProjects	AvgM
0	896999	41	finance	0.42	0.08	5	
1	331148	41	hr	0.91	0.73	5	
2	559437	36	operations	0.93	0.82	7	
3	883201	41	finance	0.03	0.53	7	
4	562242	41	finance	0.66	0.72	3	

## FEATURE ENGINEERING

### 1. Hours per Project to measure work intensity.

- Monthly Snapshot – Both AvgMonthlyHours and NumProjects represent monthly values, providing a consistent time frame for workload analysis.
- Workload per Project – Hours\_Per\_Project = AvgMonthlyHours / NumProjects measures how much time an employee spends on each project per month. High values indicate potential overwork or burnout risk.

```
In [3]: df['HoursPerProject']= df['AvgMonthlyHours']/df['NumProjects']
```

```
In [4]: # Replace any infinite values (e.g., NumProjects = 0) with median
median_hours_per_project = df['HoursPerProject'].median()
df['HoursPerProject'] = df['HoursPerProject'].replace([np.inf, -np.inf, np.nan], medi
```

### 2. Performance Ratio ~ Evaluation / Satisfaction.

- Calculates Performance\_Ratio to flag high-performing but low-satisfaction employees, replacing invalid values with the median for clean analysis.

```
In [5]: df['PerformanceRatio'] = df['LastEvaluationScore']/df['SatisfactionScore']
```

In [6]:

```
#Replace infinite or undefined values with median
median_performance_ratio = df['PerformanceRatio'].median()
df['PerformanceRatio'] = df['PerformanceRatio'].replace([np.inf, -np.inf, np.nan],
```



### 3. Tenure / Experience Feature- Categorize employees based on YearsAtCompany.

- Creates TenureCategory to group employees by experience level, enabling HR to tailor retention, promotions, and training strategies.

In [7]:

```
df['TenureCategory'] = pd.cut(
    df['YearsAtCompany'],
    bins=[0, 2, 5, 10],
    labels=['New', 'Mid', 'Experienced'],
    right=True, # includes the upper bound of each bin
    include_lowest=True
)
```

### 4. High-Risk Employee Flag-employees with high workload AND low satisfaction as "at risk".

- Flagging employees who have high workload but low satisfaction as High\_Risk\_Employee helps identify employees most likely at risk of leaving, providing actionable HR insights.

In [8]:

```
df['High_Risk_Employee'] = np.where(
    (df['HoursPerProject'] > df['HoursPerProject'].median()) &
    (df['SatisfactionScore'] < df['SatisfactionScore'].median()),
    1, 0
)
```

### 5. Attrition Insights by Department & Tenure

- Calculate average attrition by Department and TenureCategory to reveal patterns in employee turnover.
- Identify departments and tenure groups with higher attrition to guide HR retention strategies.

In [9]:

```
# Calculate average attrition by Department
dept_attrition = df.groupby('Department')['Attrition'].mean().sort_values(ascending=False)

# Calculate average attrition by TenureCategory
tenure_attrition = df.groupby('TenureCategory', observed=True)['Attrition'].mean().sort_values(ascending=False)
```



In [16]:

```
dept_attrition*100
```

```
Out[16]: Department
    hr           33.347090
    operations   33.091398
    it            32.917014
    unknown       32.889237
    finance        32.772765
    sales          32.480211
Name: Attrition, dtype: float64
```

```
In [11]: tenure_attrition
```

```
Out[11]: TenureCategory
    Mid           0.336042
    Experienced   0.329341
    New           0.320611
Name: Attrition, dtype: float64
```

```
In [12]: # New Features
print(df[['EmployeeID','AvgMonthlyHours','NumProjects','HoursPerProject',
          'SatisfactionScore','LastEvaluationScore','PerformanceRatio',
          'YearsAtCompany','TenureCategory','High_Risk_Employee','Attrition']].head(1))
```

	EmployeeID	AvgMonthlyHours	NumProjects	HoursPerProject	TenureCategory	High_Risk_Employee	Attrition
0	896999	200	5	40.000000	Experienced	0	1
1	331148	180	5	36.000000	Experienced	0	1
2	559437	200	7	28.571429	Mid	0	1
3	883201	297	7	42.428571	Experienced	1	1
4	562242	186	3	62.000000	Mid	0	0
5	538510	200	4	50.000000	Mid	0	0
6	585585	200	5	40.000000	Mid	0	0
7	689574	227	5	45.400000	Mid	0	0
8	394433	205	3	68.333333	Mid	0	0
9	314638	227	5	45.400000	Mid	0	0
	SatisfactionScore	LastEvaluationScore	PerformanceRatio	YearsAtCompany			
0	0.42	0.08	0.190476	6			
1	0.91	0.73	0.802198	7			
2	0.93	0.82	0.881720	4			
3	0.03	0.53	17.666667	6			
4	0.66	0.72	1.090909	5			
5	0.85	0.76	0.894118	6			
6	0.65	0.39	0.600000	5			
7	0.89	0.80	0.898876	6			
8	0.33	0.92	2.787879	7			
9	0.57	0.42	0.736842	6			
	TenureCategory	High_Risk_Employee	Attrition				
0	Experienced	0	1				
1	Experienced	0	1				
2	Mid	0	1				
3	Experienced	1	1				
4	Mid	0	0				
5	Experienced	0	1				
6	Mid	0	0				
7	Experienced	0	1				
8	Experienced	1	0				
9	Experienced	0	1				

In [13]:

```
#overview  
df.head()
```

Out[13]:

	EmployeeID	Age	Department	SatisfactionScore	LastEvaluationScore	NumProjects	AvgM
0	896999	41	finance	0.42	0.08	5	
1	331148	41	hr	0.91	0.73	5	
2	559437	36	operations	0.93	0.82	7	
3	883201	41	finance	0.03	0.53	7	
4	562242	41	finance	0.66	0.72	3	



In [15]:

```
# Save Feature-Enhanced Dataset  
# -----  
df.to_csv("hr_features_dataset.csv", index=False)  
print("\nFeature-engineered HR dataset saved.")
```

Feature-engineered HR dataset saved.