



DAY 11 — Pandas Advanced

Goal: **Group, merge, and transform data**

1 **groupby()** (MOST IMPORTANT)

Why **groupby** ?

Used to:

- Aggregate data
- Analyze categories
- Create features for ML

Simple Example

```
import pandas as pd

data = {
    "department": ["IT", "IT", "HR", "HR", "Sales"],
    "salary": [60000, 65000, 50000, 52000, 70000]
}

df = pd.DataFrame(data)
```

```
df.groupby("department")["salary"].mean()
```

```
# Output:
department
HR51000
IT62500
Sales70000
```

Multiple Aggregations

```
df.groupby("department")["salary"].agg(["mean","max","min"])
```

2 **merge()** (Like SQL JOIN)

Why merge?

Real data is split across files.

Example

```
employees = pd.DataFrame({  
    "emp_id": [1,2,3],  
    "name": ["Alice","Bob","Charlie"]  
})
```

```
salaries = pd.DataFrame({  
    "emp_id": [1,2,3],  
    "salary": [60000,65000,70000]  
})
```

```
merged = pd.merge(employees, salaries, on="emp_id")  
print(merged)
```

Types of joins

```
pd.merge(a, b, how="inner")  
pd.merge(a, b, how="left")  
pd.merge(a, b, how="right")  
pd.merge(a, b, how="outer")
```

3 **apply()** (Row-wise Logic)

Why apply?

When built-in functions are not enough.

Example

```
df = pd.DataFrame({  
    "score": [85,90,72]  
})  
  
df["grade"] = df["score"].apply(  
    lambda x:"Pass"if x >=80else"Fail"  
)
```

4 map() vs apply() (Simple)

Method	Used for
map()	Single column mapping
apply()	Row or column logic

Example:

```
df["score"].map(lambda x: x *1.1)
```

5 pivot_table() (Summary Tables)

```
sales = pd.DataFrame({  
    "region": ["East","East","West","West"],  
    "product": ["A","B","A","B"],  
    "revenue": [100,150,200,250]  
})
```

```
pd.pivot_table(  
    sales,  
    values="revenue",  
    index="region",  
    columns="product",  
    aggfunc="sum"  
)
```

6 ML Feature Engineering Example

```
df["salary_scaled"] = (  
    df["salary"] - df["salary"].mean()  
) / df["salary"].std()
```

This uses:

- ✓ aggregation
 - ✓ column creation
 - ✓ ML preparation
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