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1 Practical: Data integration and Data Transformation for Data Mining
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In [2]: 1 #Practical: Data integration and Data Transformation for Data Mining
2 import pandas as pd
3 import numpy as np
4
5 #sample datasets for Data integration
6
7 data_1 = {
8     'ID': [1,2,3,4],
9     'Name': ['Alice', 'Bob', 'Charlie', 'David'],
10    'Age': [25,30,35,40]
11 }
12
13 data_2 = {
14     'ID': [3,4,5,6],
15     'Gender': ['F', 'M', 'M', 'F'],
16     'Salary': [70000,80000,50000,60000]
17 }
18
19 df1 = pd.DataFrame(data_1)
20 df2 = pd.DataFrame(data_2)
```

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In [3]: 1 #data integration
2
3 #tight coupling (join datasets on common key)
4 tight_coupling= pd.merge(df1,df2, on= 'ID', how='inner')
5 print("Tight Coupling result:\n", tight_coupling)
```

Tight Coupling result:

	ID	Name	Age	Gender	Salary
0	3	Charlie	35	F	70000
1	4	David	40	M	80000

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In [5]: 1 #loose coupling (concatenate datasets)
2 loose_coupling=pd.concat([df1.set_index('ID'), df2.set_index('ID')], axis=1).reset_index()
3 print("Loose Coupling result:\n", loose_coupling)
```

Loose Coupling result:

	ID	Name	Age	Gender	Salary
0	1	Alice	25.0	NaN	NaN
1	2	Bob	30.0	NaN	NaN
2	3	Charlie	35.0	F	70000.0
3	4	David	40.0	M	80000.0
4	5	NaN	NaN	M	50000.0
5	6	NaN	NaN	F	60000.0

```
In [9]: 1 #data transformation
2 #smoothing (moving average for age)
3 loose_coupling['Smoothed_age'] = loose_coupling['Age'].rolling(window=2, min_periods=1).mean()
4 print("\n Smoothing:\n ", loose_coupling[['ID','Age','Smoothed_age']])
```

Smoothing:

	ID	Age	Smoothed_age
0	1	25.0	25.0
1	2	30.0	27.5
2	3	35.0	32.5
3	4	40.0	37.5
4	5	NaN	40.0
5	6	NaN	NaN

```
In [10]: 1 #aggregation (summarizing salary by gender)
2 aggregation = loose_coupling.groupby('Gender')['Salary'].sum().reset_index()
3 print("\n Aggregation: \n", aggregation)
```

Aggregation:

	Gender	Salary
0	F	130000.0
1	M	130000.0

```
In [14]: 1 #discretization (binning age into categories)
2 bins=[0,20,30,40,50]
3 labels=['Teen','Young Adult', 'Adult','Senior']
4 loose_coupling['Age_Group'] = pd.cut(loose_coupling['Age'], bins=bins, labels=labels)
5 print("\n Discretization: \n", loose_coupling[['ID', 'Age', 'Age_Group']])
```

Discretization:

	ID	Age	Age_Group
0	1	25.0	Young Adult
1	2	30.0	Young Adult
2	3	35.0	Adult
3	4	40.0	Adult
4	5	NaN	NaN
5	6	NaN	NaN

```
In [15]: 1 #attribute constructino (creating age salary ratio)
2 loose_coupling['Age_Salary_Ratio'] = loose_coupling['Age']/ loose_coupling['Salary']
3 print("\n Attribute construction: \n", loose_coupling[['ID', 'Age', 'Salary', 'Age_Salary_Ratio']])
```

Attribute construction:

	ID	Age	Salary	Age_Salary_Ratio
0	1	25.0	NaN	NaN
1	2	30.0	NaN	NaN
2	3	35.0	70000.0	0.0005
3	4	40.0	80000.0	0.0005
4	5	NaN	50000.0	NaN
5	6	NaN	60000.0	NaN

```
In [17]: 1 #genralization
2 loose_coupling['Age_MinMax'] = (loose_coupling['Age'] - loose_coupling['Age'].min()) / (loose_coupling['Age'].max() -loose_coupling['Age'].min())
3 print("\n Min Max Normalizarion: \n", loose_coupling[['ID', 'Age', 'Age_MinMax']])
```

Min Max Normalizarion:

	ID	Age	Age_MinMax
0	1	25.0	0.000000
1	2	30.0	0.333333
2	3	35.0	0.666667
3	4	40.0	1.000000
4	5	NaN	NaN
5	6	NaN	NaN

```
In [19]: 1 #z score normalization
2 loose_coupling['Age_Zscore'] = (loose_coupling['Age'] - loose_coupling['Age'].mean()) / loose_coupling['Age'].std()
3 print("\n z score normalization: ", loose_coupling[['ID', 'Age', 'Age_Zscore']])
```

z score normalization:

	ID	Age	Age_Zscore
0	1	25.0	-1.161895
1	2	30.0	-0.387298
2	3	35.0	0.387298
3	4	40.0	1.161895
4	5	NaN	NaN
5	6	NaN	NaN

```
In [20]: 1 #decimal scaling
2 scaling_factoe = 10 **np.ceil(np.log10(loose_coupling['Age'].abs().max()))
3 loose_coupling['Age_Decimal_Scaling'] = loose_coupling['Age'] / scaling_factoe
4 print("\n Decimal scaling:\n", loose_coupling[['ID', 'Age', 'Age_Decimal_Scaling']])
```

Decimal scaling:

	ID	Age	Age_Decimal_Scaling
0	1	25.0	0.25
1	2	30.0	0.30
2	3	35.0	0.35
3	4	40.0	0.40
4	5	NaN	NaN
5	6	NaN	NaN

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In [ ]: 1
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