

Aim: Demonstrate the working of feature construction by combining and splitting the features to extraction the information from the dataset and write a conclusion about survivals status of different age group

```
In [8]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import cross_val_score
from sklearn.linear_model import LogisticRegression
```

```
In [9]: df=pd.read_csv('DATA/train.csv')[['Age', 'Pclass', 'SibSp', 'Parch', 'Survived']]
df.head()
```

```
Out[9]:
```

	Age	Pclass	SibSp	Parch	Survived
0	22.0	3	1	0	0
1	38.0	1	1	0	1
2	26.0	3	0	0	1
3	35.0	1	1	0	1
4	35.0	3	0	0	0

```
In [10]: df.dropna(inplace=True)
```

```
In [11]: df.head()
```

```
Out[11]:
```

	Age	Pclass	SibSp	Parch	Survived
0	22.0	3	1	0	0
1	38.0	1	1	0	1
2	26.0	3	0	0	1
3	35.0	1	1	0	1
4	35.0	3	0	0	0

```
In [12]: X = df.iloc[:,0:4]
y = df.iloc[:, -1]
```

```
In [13]: X.head()
```

```
Out[13]:
```

	Age	Pclass	SibSp	Parch
0	22.0	3	1	0
1	38.0	1	1	0
2	26.0	3	0	0
3	35.0	1	1	0
4	35.0	3	0	0

```
In [15]: np.mean(cross_val_score(LogisticRegression(),X,y,scoring="accuracy",cv=20))
```

```
Out[15]: 0.6933333333333332
```

```
In [17]: X['Family_size'] = X['SibSp'] + X['Parch'] + 1
```

```
In [18]: X.head()
```

```
Out[18]:
```

	Age	Pclass	SibSp	Parch	Family_size
0	22.0	3	1	0	2
1	38.0	1	1	0	2
2	26.0	3	0	0	1
3	35.0	1	1	0	2
4	35.0	3	0	0	1

```
In [21]: cross_val_score(LogisticRegression(),X,y,scoring="accuracy",cv=20)
```

```
Out[21]: array([0.61111111, 0.63888889, 0.61111111, 0.55555556, 0.77777778,
0.55555556, 0.80555556, 0.63888889, 0.72222222, 0.72222222,
0.72222222, 0.72222222, 0.75, 0.83333333, 0.54285714,
0.88571429, 0.68571429, 0.68571429, 0.74285714, 0.65714286])
```

```
In [29]: def myfunc(num):
        if num == 1:
            return 0
        elif num > 1 and num <=4:
            return 1
        else:
            return 2
```

```
In [30]: myfunc(4)
```

Out[30]: 1

```
In [31]: X['Family_type']= X['Family_size'].apply(myfunc)
```

```
In [32]: X.head()
```

Out[32]:

	Age	Pclass	SibSp	Parch	Family_size	Family_type
0	22.0	3	1	0	2	1
1	38.0	1	1	0	2	1
2	26.0	3	0	0	1	0
3	35.0	1	1	0	2	1
4	35.0	3	0	0	1	0

```
In [33]: np.mean(cross_val_score(LogisticRegression(),X,y,scoring="accuracy",cv=20))
```

Out[33]: 0.7031746031746031

```
In [34]: df=pd.read_csv('DATA/train.csv')
```

```
In [35]: df.head()
```

Out[35]:

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female	38.0	1	0	PC 17599	71.2833	C85	C
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	S
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S

```
In [42]: df['Title']=df['Name'].str.split(', ',expand=True)[1].str.split('.',expand=True)[0]
```

```
In [45]: df[['Title','Name']]
```

Out[45]:

	Title	Name
0	Mr	Braund, Mr. Owen Harris
1	Mrs	Cumings, Mrs. John Bradley (Florence Briggs Th...
2	Miss	Heikkinen, Miss. Laina
3	Mrs	Futrelle, Mrs. Jacques Heath (Lily May Peel)
4	Mr	Allen, Mr. William Henry
...
886	Rev	Montvila, Rev. Juozas
887	Miss	Graham, Miss. Margaret Edith
888	Miss	Johnston, Miss. Catherine Helen "Carrie"
889	Mr	Behr, Mr. Karl Howell
890	Mr	Dooley, Mr. Patrick

891 rows × 2 columns

```
In [48]: (df.groupby('Title').mean()['Survived']).sort_values(False)
```

C:\Users\User23\AppData\Local\Temp\ipykernel_12820\2479167924.py:1: FutureWarning: In a future version of pandas all arguments of Series.sort_values will be keyword-only.

```
(df.groupby('Title').mean()['Survived']).sort_values(False)
```

```
Out[48]: Title
Capt      0.000000
Don        0.000000
Jonkheer   0.000000
Rev        0.000000
Mr         0.156673
Dr         0.428571
Col        0.500000
Major      0.500000
Master     0.575000
Miss       0.697802
Mrs        0.792000
Mme        1.000000
Sir        1.000000
Ms         1.000000
Lady       1.000000
Mlle       1.000000
the Countess 1.000000
Name: Survived, dtype: float64
```

```
In [49]: df['Is_married']=0
df['Is_married'].loc[df["Title"]=="Mrs"]=1
```

C:\Users\User23\AppData\Local\Temp\ipykernel_12820\2148996725.py:2: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)
df['Is_married'].loc[df["Title"]=="Mrs"]=1

```
In [50]: df['Is_married']
```

```
Out[50]: 0      0
1      1
2      0
3      1
4      0
..
886    0
887    0
888    0
889    0
890    0
Name: Is_married, Length: 891, dtype: int64
```

Conclusion: From the above experiment we conclude that the death rate of higher class people was nearly zero and deaths of nobel males was highest they sacrificed themselves to save others the rate of child and ladies was also low.

```
In [ ]:
```